

A distance metric between histograms derived from the Kolmogorov-Smirnov test statistic: specification, measures reference and example uses

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This document presents reference values for a distance metric derived from the Kolmogorov-Smirnov test statistic $D_{F,F'}$. Each measure of the $D_{F,F'}$ is a distance between two histograms. This distance is normalized by the number of observations in each sample to yield the $c = D_{F,F'} \sqrt{\frac{nn'}{n+n'}}$ statistic, which can be mapped to p-values, i.e. values for which high enough levels of significance α implies the rejection of the null hypothesis. Benchmarks for the implementation are delivered by comparing samples from known distributions. Pattern examples in real data enables further insight in the robustness and power of c .

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I. INTRODUCTION

Be F and F' two empirical cumulative distributions, where n and n' are the number of observations on each sample. The two-sample Kolmogorov-Smirnov test rejects the null hypothesis that the histograms are the outcome of the same underlying distribution if:

$$D_{F,F'} > c(\alpha) \sqrt{\frac{n+n'}{nn'}} \quad (1)$$

where $D_{F,F'} = \sup_x [F - F']$ as in Figure 1 and $c(\alpha)$ is related to the level of significance α by:

α	0.1	0.05	0.025	0.01	0.005	0.001
$c(\alpha)$	1.22	1.36	1.48	1.63	1.73	1.95

If distributions are drawn from empirical data, $D_{F,F'}$ is given as are n and n' . All terms in equation 1 are positive and $c(\alpha)$ can be isolated:

$$c(\alpha) < D_{F,F'} \sqrt{\frac{nn'}{n+n'}} = c \quad (2)$$

When c is high, low values of α favor rejecting the null hypothesis. In fact, c can be normalized to yield p-values.

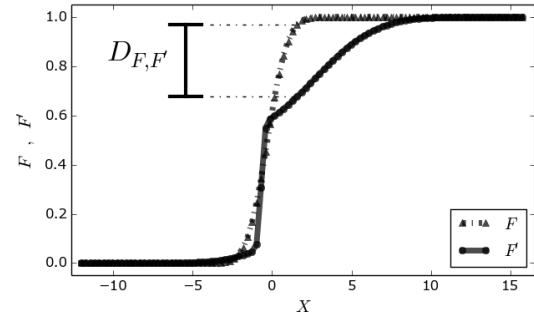


FIG. 1. The Kolmogorov-Smirnov statistic $D_{F,F'}$: the maximum difference between two cumulative distribution functions.

High values of c favor rejecting the null hypothesis. For example, if the significance level is $\alpha = 0.01$, then c greater than 1.7 implies the rejection of the null hypothesis and suggests that F and F' are outcomes of different distributions. Of core importance in this study is to regard the c statistic as a measure of distance between both distributions¹. The main contribution of the following sections is the explicit display of reference values of c from which one might derive knowledge from measures or even from a single value of c .

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A. Philosophical and technological note

Difference and equivalence is of central role in human cognition, philosophy and science. This fact is so deeply recognized that thinkers often reduce thought to classifications, e.g. through the mathematical concept of equivalence classes². Histograms are very immediate and informative roughly wherever there is a phenomenon of interest which can yield measurements. This present document should enable conclusions to be drawn about the equivalence (and difference) of the processes underlying sets of measurements for a very broad range of phenomena. The following tables also validate the mathematical framework and the software implementation.

B. Document outline

Section II exposes reference values drawn from simulations. Section III exemplifies the use of the c statistic to make sense of phenomena. Section IV holds final remarks with directions to software and data.

II. REFERENCES THROUGH SIMULATIONS

Values of c are given for simulations involving normal, uniform, Weibull and power function distributions. The rendering of this article is automated to ease changes in the settings with which the results are reported. The number of comparisons is $N_c = 100$, each with the sample sizes of $n = 1000$ and $n' = 1000$. Each histogram have $N_b = 300$ equally spaced bins.

A. When the null hypothesis is true

If the null hypothesis is true, the number of rejections of the null hypothesis ($c > c(\alpha)$) in N_c comparisons should not exceed αN_c . To verify this, let $C = \{c_i\}$ be a set of c measures, and $C(\alpha) = \{c : c > c(\alpha)\}$. Be $|C(\alpha)|$ the cardinality of $C(\alpha)$, i.e. the number of comparisons in which the two-sample Kolmogorov-Smirnov test rejects the null hypothesis for a given α . This section reports that $|C(\alpha)|$ very rarely exceeds αN_c , for all probability distributions and settings. Also important are that $c > c(\alpha)$ in many cases and that the tabulated α values are also good estimates of the upper limit of the frequency of such an event.

αN_c	α	$c(\alpha)$	$ C_1(\alpha) $	$ C_2(\alpha) $	$ C_3(\alpha) $
10.0	0.100	1.22	8	7	5
5.0	0.050	1.36	1	2	2
2.5	0.025	1.48	1	0	1
1.0	0.010	1.63	1	0	0
0.5	0.005	1.73	1	0	0
0.1	0.001	1.95	0	0	0

TABLE I. The theoretical maximum number αN_c of rejections of the null hypothesis for significance levels α . The c_1 values were calculated using simulations of normal distributions with $\mu = 0$ and $\sigma = 1$. The c_2 values were calculated using simulations of normal distributions with $\mu = 3$ and $\sigma = 2$. The c_3 values were calculated using simulations of normal distributions with $\mu = 6$ and $\sigma = 3$. Over all N_c comparisons, $\mu(c_1) = 0.7775$ and $\sigma(c_1) = 0.2573$, $\mu(c_2) = 0.7976$ and $\sigma(c_2) = 0.2388$, $\mu(c_3) = 0.7596$ and $\sigma(c_3) = 0.2292$.

αN_c	α	$c(\alpha)$	$ C_1(\alpha) $	$ C_2(\alpha) $	$ C_3(\alpha) $
10.0	0.100	1.22	9	9	4
5.0	0.050	1.36	5	4	0
2.5	0.025	1.48	2	1	0
1.0	0.010	1.63	0	0	0
0.5	0.005	1.73	0	0	0
0.1	0.001	1.95	0	0	0

TABLE II. The theoretical maximum number αN_c of rejections of the null hypothesis for critical values of α . The c_1 values were calculated using simulations of uniform distributions within $[0, 1)$. The c_2 values were calculated using simulations of uniform distributions within $[2, 6)$. The c_3 values were calculated using simulations of uniform distributions with $\mu = 4$ and $\sigma = 10$. Over all N_c comparisons, $\mu(c_1) = 0.8674$ and $\sigma(c_1) = 0.2468$, $\mu(c_2) = 0.8298$ and $\sigma(c_2) = 0.2608$, $\mu(c_3) = 0.7983$ and $\sigma(c_3) = 0.2205$.

αN_c	α	$c(\alpha)$	$ C_1(\alpha) $	$ C_2(\alpha) $	$ C_3(\alpha) $	$ C_4(\alpha) $
10.0	0.100	1.22	0	9	5	9
5.0	0.050	1.36	0	1	3	3
2.5	0.025	1.48	0	0	1	1
1.0	0.010	1.63	0	0	1	0
0.5	0.005	1.73	0	0	1	0
0.1	0.001	1.95	0	0	0	0

TABLE III. The theoretical maximum number αN_c of rejections of the null hypothesis for critical values of α . The c_1 values were calculated using simulations of 1-parameter Weibull distributions with $a = 0.1$. The c_2 values were calculated using simulations of 1-parameter Weibull distributions with $a = 2$. The c_3 values were calculated using simulations of 1-parameter Weibull distributions with $a = 4$. Over all N_c comparisons, The N_o values of c_4 were calculated using simulations of 1-parameter Weibull distributions with $a = 6$. Over all N_c comparisons, $\mu(c_1) = 0.1107$ and $\sigma(c_1) = 0.0652$, $\mu(c_2) = 0.8079$ and $\sigma(c_2) = 0.2417$, $\mu(c_3) = 0.7775$ and $\sigma(c_3) = 0.2404$, $\mu(c_4) = 0.8209$ and $\sigma(c_4) = 0.2389$.

αN_c	α	$c(\alpha)$	$ C_1(\alpha) $	$ C_2(\alpha) $	$ C_3(\alpha) $	$ C_4(\alpha) $	$ C_5(\alpha) $
10.0	0.100	1.22	13	10	10	10	9
5.0	0.050	1.36	8	3	7	5	7
2.5	0.025	1.48	5	2	3	2	2
1.0	0.010	1.63	3	0	1	1	1
0.5	0.005	1.73	1	0	0	1	1
0.1	0.001	1.95	0	0	0	0	0

TABLE IV. The theoretical maximum number αN_c of rejections of the null hypothesis for critical values of α . The c_1 values were calculated using simulations of power functions distributions with $a = 0.3$. The c_2 values were calculated using simulations of power functions distributions with $a = 1$. The c_3 values were calculated using simulations of power functions distributions with $a = 2$. The c_4 values were calculated using simulations of power functions distributions with $a = 3$. The c_5 values were calculated using simulations of power functions distributions with $a = 4$. Over all N_c comparisons, $\mu(c_1) = 0.8379$ and $\sigma(c_1) = 0.3099$, $\mu(c_2) = 0.8392$ and $\sigma(c_2) = 0.2557$, $\mu(c_3) = 0.8642$ and $\sigma(c_3) = 0.2747$. $\mu(c_4) = 0.8383$ and $\sigma(c_4) = 0.2724$. $\mu(c_5) = 0.7920$ and $\sigma(c_5) = 0.2779$.

B. When the null hypothesis is false

The null hypothesis is always false for a sufficiently small significance level α . In this section, each table holds a set comparisons between two samples: one sample is generated through a fixed distribution while the other sample is modified in each comparison. The comparison is repeated N_c times. The measures on c chosen to report the results are: the mean $\mu(c)$, the standard deviation $\sigma(c)$, the median $m(c)$, the fraction $\overline{C(\alpha)} = \frac{|C(\alpha)|}{N_c}$ of rejection of the null hypothesis given the significance level α . The null hypothesis is true in the boldface lines. Let D be the KS statistic when sample size goes to infinity. Notably, high values of c yield smaller $|D - D_{F,F'}|$, i.e. more accurate estimates of D through $D_{F,F'}$.

σ	$\mu(c)$	$\sigma(c)$	$\min(c)$	$\max(c)$	D	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
0.5	3.978	0.246	3.309,3.354,3.488	4.427,4.517,4.763	0.161	0.178	0.011	1.000	1.000	1.000	1.000	1.000	1.000
0.6	3.084	0.288	2.549,2.571,2.616	3.690,3.734,3.935	0.121	0.138	0.013	1.000	1.000	1.000	1.000	1.000	1.000
0.7	2.296	0.222	1.856,1.878,1.901	2.750,2.750,2.840	0.085	0.103	0.010	1.000	1.000	1.000	1.000	1.000	0.950
0.8	1.614	0.278	0.984,1.140,1.140	2.370,2.370,2.437	0.054	0.072	0.012	0.940	0.870	0.660	0.440	0.300	0.100
0.9	1.123	0.273	0.492,0.559,0.626	1.610,1.744,1.901	0.025	0.050	0.012	0.340	0.230	0.120	0.020	0.020	0.000
1.0	0.851	0.284	0.425,0.447,0.447	1.521,1.588,1.856	0.000	0.038	0.013	0.110	0.060	0.030	0.010	0.010	0.000
1.1	1.064	0.258	0.604,0.671,0.671	1.632,1.655,1.923	0.023	0.048	0.012	0.260	0.160	0.060	0.030	0.010	0.000
1.2	1.447	0.279	0.693,0.872,0.939	1.990,2.057,2.057	0.044	0.065	0.012	0.790	0.650	0.400	0.270	0.170	0.040
1.3	1.825	0.262	1.230,1.230,1.275	2.348,2.415,2.750	0.063	0.082	0.012	1.000	0.970	0.900	0.820	0.650	0.290
1.4	2.209	0.292	1.699,1.722,1.766	2.840,2.862,2.862	0.081	0.099	0.013	1.000	1.000	1.000	1.000	0.980	0.790
1.5	2.564	0.301	1.901,1.990,2.012	3.153,3.220,3.309	0.097	0.115	0.013	1.000	1.000	1.000	1.000	1.000	0.990
1.6	2.895	0.279	2.214,2.326,2.348	3.444,3.488,3.578	0.112	0.129	0.012	1.000	1.000	1.000	1.000	1.000	1.000
1.7	3.152	0.290	2.482,2.616,2.616	3.868,3.868,3.891	0.125	0.141	0.013	1.000	1.000	1.000	1.000	1.000	1.000
1.8	3.431	0.266	2.750,2.795,2.974	4.003,4.003,4.025	0.138	0.153	0.012	1.000	1.000	1.000	1.000	1.000	1.000
1.9	3.742	0.293	3.108,3.153,3.198	4.316,4.494,4.763	0.150	0.167	0.013	1.000	1.000	1.000	1.000	1.000	1.000
2.0	3.971	0.269	3.354,3.466,3.511	4.472,4.494,4.584	0.161	0.178	0.012	1.000	1.000	1.000	1.000	1.000	1.000

TABLE V. Measurements of c through simulations with normal distributions. One normal distribution is fixed, with $\mu = 0$ and $\sigma = 1$, and compared against normal distributions with $\mu = 0$ and different values of σ .

μ	$\mu(c)$	$\sigma(c)$	$\min(c)$	$\max(c)$	D	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
0.0	0.829	0.271	0.402,0.402,0.402	1.565,1.677,1.677	0.000	0.037	0.012	0.080	0.050	0.040	0.020	0.000	0.000
0.1	1.357	0.414	0.537,0.559,0.581	2.102,2.303,2.415	0.040	0.061	0.019	0.540	0.490	0.340	0.260	0.210	0.110
0.2	2.098	0.451	0.939,1.096,1.118	3.130,3.153,3.198	0.080	0.094	0.020	0.970	0.950	0.910	0.870	0.790	0.640
0.3	3.083	0.399	2.124,2.147,2.191	3.846,3.868,3.935	0.119	0.138	0.018	1.000	1.000	1.000	1.000	1.000	1.000
0.4	3.820	0.477	2.348,2.996,3.130	4.785,4.942,5.411	0.159	0.171	0.021	1.000	1.000	1.000	1.000	1.000	1.000
0.5	4.637	0.449	3.645,3.757,3.801	5.545,5.613,5.680	0.197	0.207	0.020	1.000	1.000	1.000	1.000	1.000	1.000
0.6	5.548	0.483	4.383,4.472,4.584	6.373,6.507,6.507	0.236	0.248	0.022	1.000	1.000	1.000	1.000	1.000	1.000
0.7	6.362	0.415	5.009,5.344,5.568	7.178,7.200,7.357	0.274	0.285	0.019	1.000	1.000	1.000	1.000	1.000	1.000
0.8	7.158	0.441	6.104,6.149,6.306	8.027,8.072,8.408	0.311	0.320	0.020	1.000	1.000	1.000	1.000	1.000	1.000
0.9	7.888	0.419	6.820,6.909,7.133	8.765,8.788,8.967	0.347	0.353	0.019	1.000	1.000	1.000	1.000	1.000	1.000
1.0	8.811	0.413	7.759,7.782,7.938	9.481,9.615,9.839	0.383	0.394	0.018	1.000	1.000	1.000	1.000	1.000	1.000

TABLE VI. Measurements of c through simulations with normal distributions. One normal distribution is fixed, with $\mu = 0$ and $\sigma = 1$, and compared against normal distributions with different values of μ and fixed $\sigma = 1$.

b	$\mu(c)$	$\sigma(c)$	$\min(c)$	$\max(c)$	D	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
0.7	6.750	0.377	5.858,5.970,6.037	7.536,7.580,7.647	0.300	0.302	0.017	1.000	1.000	1.000	1.000	1.000	1.000
0.75	5.645	0.303	4.964,5.009,5.076	6.239,6.373,6.619	0.250	0.252	0.014	1.000	1.000	1.000	1.000	1.000	1.000
0.8	4.573	0.287	3.824,3.958,4.047	5.121,5.165,5.299	0.200	0.205	0.013	1.000	1.000	1.000	1.000	1.000	1.000
0.85	3.457	0.234	2.840,2.907,3.019	3.846,4.092,4.114	0.150	0.155	0.010	1.000	1.000	1.000	1.000	1.000	1.000
0.9	2.362	0.251	1.811,1.834,1.901	2.885,2.885,3.019	0.100	0.106	0.011	1.000	1.000	1.000	1.000	1.000	0.930
0.95	1.352	0.255	0.939,0.962,0.962	2.080,2.147,2.214	0.050	0.060	0.011	0.690	0.400	0.270	0.120	0.080	0.030
1.0	0.797	0.235	0.358,0.425,0.447	1.431,1.453,1.476	0.000	0.036	0.011	0.060	0.030	0.000	0.000	0.000	0.000
1.05	1.343	0.310	0.805,0.894,0.917	2.147,2.281,2.303	0.048	0.060	0.014	0.550	0.390	0.250	0.180	0.140	0.050
1.1	2.191	0.295	1.565,1.588,1.655	2.706,2.706,2.885	0.091	0.098	0.013	1.000	1.000	1.000	0.980	0.940	0.760
1.15	3.032	0.294	2.370,2.393,2.504	3.645,3.779,3.935	0.130	0.136	0.013	1.000	1.000	1.000	1.000	1.000	1.000
1.2	3.822	0.279	2.885,3.063,3.242	4.360,4.383,4.562	0.167	0.171	0.012	1.000	1.000	1.000	1.000	1.000	1.000
1.25	4.532	0.294	3.801,3.824,3.958	5.188,5.188,5.456	0.200	0.203	0.013	1.000	1.000	1.000	1.000	1.000	1.000
1.3	5.258	0.310	4.383,4.427,4.539	5.903,5.993,6.104	0.231	0.235	0.014	1.000	1.000	1.000	1.000	1.000	1.000
1.35	5.799	0.287	5.098,5.188,5.188	6.261,6.261,6.283	0.259	0.259	0.013	1.000	1.000	1.000	1.000	1.000	1.000

TABLE VII. Measurements of c through simulations with uniform distributions. One uniform distribution has the fixed domain $[0, 1]$. The other uniform distribution in each comparison is also centered around 0.5, but spread over $b = b_u - b_l$ there b_l and b_u are the lower and upper boudaries.

μ	$\mu(c)$	$\sigma(c)$	min(c)	max(c)	D	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
0.5	0.793	0.253	0.402,0.425,0.425	1.431,1.521,1.610	0.000	0.035	0.011	0.080	0.030	0.020	0.000	0.000	0.000
0.55	1.701	0.320	1.096,1.118,1.207	2.437,2.549,2.639	0.050	0.076	0.014	0.970	0.850	0.710	0.590	0.450	0.200
0.6	2.813	0.320	2.035,2.191,2.191	3.511,3.600,3.757	0.100	0.126	0.014	1.000	1.000	1.000	1.000	1.000	1.000
0.65	3.972	0.347	3.309,3.309,3.354	4.606,4.629,4.808	0.150	0.178	0.016	1.000	1.000	1.000	1.000	1.000	1.000
0.7	5.073	0.343	4.316,4.405,4.427	5.858,5.858,5.948	0.200	0.227	0.015	1.000	1.000	1.000	1.000	1.000	1.000
0.75	6.205	0.369	5.389,5.613,5.635	7.066,7.290,7.491	0.250	0.278	0.017	1.000	1.000	1.000	1.000	1.000	1.000
0.8	7.238	0.354	6.596,6.641,6.663	8.117,8.117,8.184	0.300	0.324	0.016	1.000	1.000	1.000	1.000	1.000	1.000
0.85	8.337	0.357	7.379,7.647,7.670	9.123,9.190,9.436	0.350	0.373	0.016	1.000	1.000	1.000	1.000	1.000	1.000
0.9	9.425	0.296	8.743,8.855,8.855	10.018,10.085,10.331	0.400	0.422	0.013	1.000	1.000	1.000	1.000	1.000	1.000
0.95	10.520	0.355	9.749,9.883,9.906	11.136,11.158,11.471	0.450	0.470	0.016	1.000	1.000	1.000	1.000	1.000	1.000
1.0	11.661	0.355	10.867,10.957,10.979	12.567,12.611,12.679	0.500	0.521	0.016	1.000	1.000	1.000	1.000	1.000	1.000
1.05	12.722	0.317	11.784,11.918,12.008	13.327,13.349,13.483	0.550	0.569	0.014	1.000	1.000	1.000	1.000	1.000	1.000
1.1	13.856	0.311	13.126,13.215,13.260	14.445,14.467,14.669	0.600	0.620	0.014	1.000	1.000	1.000	1.000	1.000	1.000

TABLE VIII. Measurements of c through simulations with uniform distributions. One uniform distribution has the fixed domain $[0, 1)$. The other uniform distribution in each comparison have varied mean values but always spread over a fixed $b = b_u - b_l$ there b_l and b_u are the lower and upper boudaries.

a	$\mu(c)$	$\sigma(c)$	min(c)	max(c)	D	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
0.7	4.604	0.394	3.801,3.868,3.891	5.344,5.501,5.545	0.201	0.206	0.018	1.000	1.000	1.000	1.000	1.000	1.000
0.9	3.310	0.393	2.393,2.460,2.571	4.137,4.181,4.226	0.136	0.148	0.018	1.000	1.000	1.000	1.000	1.000	1.000
1.1	2.233	0.400	1.431,1.498,1.543	3.130,3.153,3.511	0.083	0.100	0.018	1.000	1.000	0.990	0.940	0.880	0.760
1.3	1.336	0.341	0.693,0.738,0.805	1.990,2.147,2.415	0.039	0.060	0.015	0.610	0.410	0.310	0.200	0.150	0.050
1.5	0.881	0.274	0.470,0.470,0.492	1.632,1.655,1.655	0.000	0.039	0.012	0.080	0.080	0.060	0.030	0.000	0.000
1.7	1.251	0.319	0.648,0.738,0.760	2.102,2.102,2.124	0.034	0.056	0.014	0.520	0.370	0.210	0.120	0.090	0.030
1.9	1.762	0.344	0.984,0.984,1.185	2.571,2.616,2.862	0.064	0.079	0.015	0.960	0.900	0.810	0.660	0.450	0.260
2.1	2.271	0.345	1.409,1.476,1.543	2.952,3.220,3.421	0.090	0.102	0.015	1.000	1.000	0.980	0.950	0.930	0.840
2.3	2.851	0.378	1.990,2.124,2.191	3.600,3.846,3.913	0.114	0.127	0.017	1.000	1.000	1.000	1.000	1.000	1.000
2.5	3.271	0.371	2.348,2.393,2.594	4.047,4.092,4.114	0.136	0.146	0.017	1.000	1.000	1.000	1.000	1.000	1.000
2.7	3.695	0.367	2.929,3.130,3.130	4.517,4.517,4.808	0.155	0.165	0.016	1.000	1.000	1.000	1.000	1.000	1.000
2.9	4.170	0.379	3.309,3.354,3.399	5.009,5.076,5.098	0.173	0.186	0.017	1.000	1.000	1.000	1.000	1.000	1.000

TABLE IX. Measurements of c through simulations with 1-parameter Weibull distributions. One Weibull distribution has the fixed shape parameter $a = 1.5$. The other Weibull distribution in each comparison has varied values of a .

a	$\mu(c)$	$\sigma(c)$	min(c)	max(c)	D	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
0.7	6.282	0.402	4.919,5.299,5.501	6.909,6.999,7.021	0.274	0.281	0.018	1.000	1.000	1.000	1.000	1.000	1.000
0.9	4.445	0.452	3.511,3.600,3.622	5.344,5.367,5.590	0.186	0.199	0.020	1.000	1.000	1.000	1.000	1.000	1.000
1.1	2.818	0.443	1.588,1.744,1.945	3.734,3.846,4.137	0.114	0.126	0.020	1.000	1.000	1.000	0.990	0.990	0.970
1.3	1.536	0.407	0.783,0.827,0.894	2.415,2.437,2.549	0.053	0.069	0.018	0.750	0.650	0.540	0.350	0.300	0.170
1.5	0.776	0.237	0.425,0.447,0.470	1.409,1.409,1.565	0.000	0.035	0.011	0.060	0.040	0.010	0.000	0.000	0.000
1.7	1.499	0.377	0.648,0.738,0.850	2.281,2.326,2.370	0.046	0.067	0.017	0.740	0.660	0.510	0.380	0.300	0.110
1.9	2.246	0.400	0.939,1.386,1.521	2.952,3.063,3.309	0.087	0.100	0.018	0.990	0.990	0.980	0.940	0.870	0.780
2.1	3.051	0.395	2.393,2.393,2.415	3.846,3.891,3.913	0.123	0.136	0.018	1.000	1.000	1.000	1.000	1.000	1.000
2.3	3.710	0.442	2.683,2.795,2.907	4.696,4.718,4.785	0.156	0.166	0.020	1.000	1.000	1.000	1.000	1.000	1.000
2.5	4.415	0.422	3.019,3.175,3.287	5.121,5.188,5.210	0.186	0.197	0.019	1.000	1.000	1.000	1.000	1.000	1.000

TABLE X. Measurements of c through simulations with power function distributions. One power distribution has the fixed exponent parameter $1 - a = 2.5$. The other power function distribution in each comparison has varied values of a .

C. Changing the sample sizes

Changing the number of elements in each sample changes the value of the c statistic. This section is dedicated to tables in which the c statistic is given for two samples of varied sizes but with fixed underlying distributions. If a same value ϵ is changed of the mean μ or the standard deviation, the first yields greater values of the c statistic. In summary, raising the sample sizes raises c and gives a value of $D_{F,F'}$ closer to D (the maximum difference between the theoretical cumulative distributions).

$n = n'$	$\mu(c)$	$\sigma(c)$	$m(c)$	$\min(c)$	$\max(c)$	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
100	0.912	0.291	0.919	0.424,0.424,0.424	1.485,1.556,1.768	0.129	0.041	0.140	0.060	0.040	0.010	0.010	0.000
1000	1.283	0.364	1.286	0.626,0.648,0.648	2.080,2.124,2.147	0.057	0.016	0.540	0.390	0.320	0.150	0.100	0.060
10000	3.113	0.462	3.051	2.072,2.128,2.185	4.080,4.278,4.462	0.044	0.007	1.000	1.000	1.000	1.000	1.000	1.000
100000	9.126	0.485	9.076	7.978,8.003,8.128	10.156,10.194,10.252	0.041	0.002	1.000	1.000	1.000	1.000	1.000	1.000

TABLE XI. Measurements of c through simulations with fixed normal distributions but different number of samples. One normal distribution has $\mu = 0$ and $\sigma = 1$. The other normal distribution have $\mu = 0.1$ and $\sigma = 1$. The KS statistic of these distributions converges to 0.0399 as sample sizes increases.

$n = n'$	$\mu(c)$	$\sigma(c)$	$m(c)$	$\min(c)$	$\max(c)$	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
100	0.911	0.251	0.849	0.424,0.495,0.495	1.414,1.626,1.697	0.129	0.035	0.130	0.050	0.020	0.010	0.000	0.000
1000	1.466	0.260	1.431	0.917,1.051,1.051	1.990,2.080,2.214	0.066	0.012	0.820	0.620	0.420	0.270	0.180	0.050
10000	3.467	0.243	3.465	2.878,2.970,2.991	3.946,4.080,4.094	0.049	0.003	1.000	1.000	1.000	1.000	1.000	1.000
100000	10.129	0.253	10.125	9.595,9.595,9.631	10.713,10.735,10.896	0.045	0.001	1.000	1.000	1.000	1.000	1.000	1.000

TABLE XII. Measurements of c through simulations with fixed normal distributions but different number of samples. One normal distribution has $\mu = 0$ and $\sigma = 1$. The other normal distribution have $\mu = 0$ and $\sigma = 1.2$. The KS statistic of these distributions converges to 0.0440 as sample sizes increases.

$n = n'$	$\mu(c)$	$\sigma(c)$	$m(c)$	$\min(c)$	$\max(c)$	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
100	0.969	0.295	0.919	0.424,0.495,0.566	1.556,1.768,1.838	0.137	0.042	0.220	0.130	0.100	0.010	0.010	0.000
1000	1.692	0.311	1.677	1.140,1.185,1.207	2.326,2.460,2.661	0.076	0.014	0.970	0.870	0.710	0.560	0.420	0.200
10000	4.156	0.363	4.108	3.345,3.528,3.528	5.077,5.084,5.233	0.059	0.005	1.000	1.000	1.000	1.000	1.000	1.000
100000	11.741	0.255	11.713	11.285,11.301,11.328	12.220,12.274,12.368	0.053	0.001	1.000	1.000	1.000	1.000	1.000	1.000

TABLE XIII. Measurements of c through simulations with fixed uniform distributions but different number of samples. One distribution is uniform in $[0,1]$. The other distribution is uniform in $[0.05,1.05]$. The KS statistic of these distributions converges to 0.0500 as sample sizes increases.

$n = n'$	$\mu(c)$	$\sigma(c)$	$m(c)$	$\min(c)$	$\max(c)$	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
100	1.060	0.277	1.061	0.566,0.636,0.636	1.768,1.768,1.838	0.150	0.039	0.260	0.130	0.080	0.040	0.030	0.000
1000	2.050	0.188	2.057	1.677,1.699,1.744	2.482,2.594,2.661	0.092	0.008	1.000	1.000	1.000	1.000	0.980	0.670
10000	6.022	0.141	6.032	5.685,5.720,5.756	6.300,6.350,6.371	0.085	0.002	1.000	1.000	1.000	1.000	1.000	1.000
100000	18.742	0.169	18.737	18.329,18.385,18.418	19.069,19.071,19.103	0.084	0.001	1.000	1.000	1.000	1.000	1.000	1.000

TABLE XIV. Measurements of c through simulations with fixed uniform distributions but different number of samples. One distribution is uniform in $[0,1]$. The other distribution is uniform in $[-0.1,1.1]$. The KS statistic of these distributions converges to 0.0833 as sample sizes increases.

$n = n'$	$\mu(c)$	$\sigma(c)$	$m(c)$	$\min(c)$	$\max(c)$	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
100	0.824	0.231	0.778	0.424,0.424,0.495	1.273,1.344,1.414	0.117	0.033	0.060	0.010	0.000	0.000	0.000	0.000
1000	1.211	0.276	1.174	0.760,0.783,0.783	1.878,1.968,1.990	0.054	0.012	0.440	0.280	0.170	0.070	0.050	0.020
10000	2.624	0.358	2.588	1.824,1.945,2.015	3.458,3.642,3.684	0.037	0.005	1.000	1.000	1.000	1.000	1.000	0.980
100000	7.699	0.409	7.744	6.538,6.657,6.878	8.573,8.580,8.765	0.034	0.002	1.000	1.000	1.000	1.000	1.000	1.000

TABLE XV. Measurements of c through simulations with fixed Weibull distributions but different number of samples. One distribution has shape parameter $a = 1.5$. The other distribution has $a = 1.7$. The KS statistic of these distributions converges to 0.0338 as sample sizes increases.

$n = n'$	$\mu(c)$	$\sigma(c)$	$m(c)$	$\min(c)$	$\max(c)$	$\mu(D_{F,F'})$	$\sigma(D_{F,F'})$	$C(0.1)$	$C(0.05)$	$C(0.025)$	$C(0.01)$	$C(0.005)$	$C(0.001)$
100	0.853	0.307	0.778	0.354,0.424,0.495	1.556,1.909,1.909	0.121	0.043	0.110	0.080	0.060	0.020	0.020	0.000
1000	1.472	0.385	1.453	0.581,0.604,0.648	2.191,2.303,2.437	0.066	0.017	0.740	0.600	0.470	0.330	0.270	0.130
10000	3.494	0.433	3.500	2.574,2.666,2.673	4.342,4.469,4.518	0.049	0.006	1.000	1.000	1.000	1.000	1.000	1.000
100000	10.506	0.420	10.505	9.539,9.682,9.700	11.259,11.326,11.536	0.047	0.002	1.000	1.000	1.000	1.000	1.000	1.000

TABLE XVI. Measurements of c through simulations with fixed power distributions but different number of samples. One distribution has shape parameter $a=1.5$. The other distribution has $a=1.7$. The KS statistic of these distributions converges to 0.0460 as sample sizes increases.

III. EXAMPLE USES IN EMPIRICAL DATA

This section presents immediate results drawn from the statistic c when observed in real samples. The sample choices are arbitrary.

A. Text

This section exemplifies the use of c in the detection of similarity between texts. Each text X was divided in two halves $X1$ and $X2$. The set of known English words were considered as were the set of stopwords (words with reduced meaning such as prepositions and articles). Only the number of letters in each word was measured. Three approaches were chosen: 1) the text was partitioned into 1000 pieces of equal number of characters, the mean of the word size of each piece is an element of the sample; 2) the text was partitioned into 1000 pieces of equal number of characters, the standard deviation of the word size is an element of the sample; 3) each word size is an element of the sample. This last case yields a discrete probability distribution, which was approximated as a continuous variable and gave the greatest sensibility to text differences. The overall result is the same: smaller differences between parts of the same text. Notice that the c is often high within a same book. The Bible was the only book where the c statistic is higher between the halves than between the whole text and the halves. This might be due to the differences of the Old and New Testaments.

label	description	chars	tokens	sentences	$ kw $	$\mu(kw)$	$\sigma(kw)$	$ sw $	$\mu(sw)$	$\sigma(sw)$
H,H1,H2	Hamlet by Shakespeare	162881	37360	3106	16722	3.549	1.762	9908	2.721	1.011
B,B1,B2	King James Version of the Holly Bible	4332554	1010654	30103	492901	3.745	1.711	289244	2.927	1.044
M,M1,M2	Moby Dick by Herman Melville	1242990	260819	10059	136008	4.105	2.184	75385	2.847	1.096
E,E1,E2	Esaú e Jacó from Machado de Assis	355706	88472	3822	13984	2.186	1.376	3535	1.486	0.502

TABLE XVII. General description of the texts used to exemplify the use of the c statistic. Individual values of number of characters, tokens, sentences give context. Mean and standard deviation of the size of known words kw and of the stopwords st are used in next tables. Numbers in the labels indicate first and second half of the corresponding text in the next tables.

	H	H1	H2	B	B1	B2	M	M1	M2	E	E1	E2
H	0.000 0.000	3.421 0.153	3.690 0.165	11.560 0.517	9.749 0.436	12.455 0.557	14.847 0.664	13.975 0.625	13.953 0.624	20.321 0.909	19.574 0.875	19.185 0.858
H1	3.421 0.153	0.000 0.000	0.693 0.031	12.790 0.572	11.583 0.518	13.260 0.593	14.602 0.653	13.953 0.624	13.864 0.620	18.464 0.826	17.600 0.787	17.307 0.774
H2	3.690 0.165	0.693 0.031	0.000 0.000	13.349 0.597	11.717 0.524	13.797 0.617	15.093 0.675	14.378 0.643	14.400 0.644	18.595 0.832	17.712 0.792	17.419 0.779
B	11.560 0.517	12.790 0.572	13.349 0.597	0.000 0.000	3.980 0.178	2.929 0.131	15.474 0.692	13.685 0.612	14.065 0.629	22.271 0.996	21.888 0.979	21.802 0.975
B1	9.749 0.436	11.583 0.518	11.717 0.524	3.980 0.178	0.000 0.000	5.993 0.268	15.451 0.691	13.819 0.618	14.177 0.634	22.159 0.991	21.708 0.971	21.667 0.969
B2	12.455 0.557	13.260 0.593	13.797 0.617	2.929 0.131	5.993 0.268	0.000 0.000	14.020 0.627	12.276 0.549	12.321 0.551	22.271 0.996	21.933 0.981	21.846 0.977
M	14.847 0.664	14.602 0.653	15.093 0.675	15.474 0.692	15.451 0.691	14.020 0.627	0.000 0.000	1.923 0.086	1.789 0.080	22.271 0.996	21.821 0.976	21.779 0.974
M1	13.975 0.625	13.953 0.624	14.378 0.643	13.685 0.612	13.819 0.618	12.276 0.549	1.923 0.086	0.000 0.000	1.029 0.046	22.159 0.991	21.686 0.970	21.645 0.968
M2	13.953 0.624	13.864 0.620	14.400 0.644	14.065 0.629	14.177 0.634	12.321 0.551	1.789 0.080	1.029 0.046	0.000 0.000	22.181 0.992	21.708 0.971	21.690 0.970
E	20.321 0.909	18.464 0.826	18.595 0.832	22.271 0.996	22.159 0.991	22.271 0.996	22.271 0.996	22.159 0.991	22.181 0.992	0.000 0.000	3.236 0.145	2.495 0.112
E1	19.574 0.875	17.600 0.787	17.712 0.792	21.888 0.979	21.708 0.971	21.933 0.981	21.821 0.976	21.686 0.970	21.708 0.971	3.236 0.145	0.000 0.000	1.395 0.062
E2	19.185 0.858	17.307 0.774	17.419 0.779	21.802 0.975	21.667 0.969	21.846 0.977	21.779 0.974	21.645 0.968	21.690 0.970	2.495 0.112	1.395 0.062	0.000 0.000

TABLE XVIII. Values of c for histograms drawn from mean of the sizes of the known words.

	H	H1	H2	B	B1	B2	M	M1	M2	E	E1	E2
H	0.000	2.817	3.868	6.641	5.098	7.491	12.924	11.270	11.203	8.592	11.537	10.554
	0.000	0.126	0.173	0.297	0.228	0.335	0.578	0.504	0.501	0.384	0.516	0.472
H1	2.817	0.000	1.453	8.810	7.312	9.705	13.707	12.634	12.388	6.171	9.050	8.318
	0.126	0.000	0.065	0.394	0.327	0.434	0.613	0.565	0.554	0.276	0.405	0.372
H2	3.868	1.453	0.000	10.308	8.430	10.867	14.646	13.349	13.394	5.292	8.480	7.826
	0.173	0.065	0.000	0.461	0.377	0.486	0.655	0.597	0.599	0.237	0.379	0.350
B	6.641	8.810	10.308	0.000	3.958	2.616	16.122	13.797	14.244	14.641	16.256	15.295
	0.297	0.394	0.461	0.000	0.177	0.117	0.721	0.617	0.637	0.655	0.727	0.684
B1	5.098	7.312	8.430	3.958	0.000	5.970	16.703	14.557	14.825	13.320	15.354	14.378
	0.228	0.327	0.377	0.177	0.000	0.267	0.747	0.651	0.663	0.596	0.687	0.643
B2	7.491	9.705	10.867	2.616	5.970	0.000	14.266	12.254	12.746	15.247	16.529	15.809
	0.335	0.434	0.486	0.117	0.267	0.000	0.638	0.548	0.570	0.682	0.739	0.707
M	12.924	13.707	14.646	16.122	16.703	14.266	0.000	2.326	1.923	17.964	18.512	17.933
	0.578	0.613	0.655	0.721	0.747	0.638	0.000	0.104	0.086	0.803	0.828	0.802
M1	11.270	12.634	13.349	13.797	14.557	12.254	2.326	0.000	0.626	17.155	17.967	17.419
	0.504	0.565	0.597	0.617	0.651	0.548	0.104	0.000	0.028	0.767	0.804	0.779
M2	11.203	12.388	13.394	14.244	14.825	12.746	1.923	0.626	0.000	16.932	17.945	17.285
	0.501	0.554	0.599	0.637	0.663	0.570	0.086	0.028	0.000	0.757	0.803	0.773
E	8.592	6.171	5.292	14.641	13.320	15.247	17.964	17.155	16.932	0.000	4.389	4.031
	0.384	0.276	0.237	0.655	0.596	0.682	0.803	0.767	0.757	0.000	0.196	0.180
E1	11.537	9.050	8.480	16.256	15.354	16.529	18.512	17.967	17.945	4.389	0.000	1.138
	0.516	0.405	0.379	0.727	0.687	0.739	0.828	0.804	0.803	0.196	0.000	0.051
E2	10.554	8.318	7.826	15.295	14.378	15.809	17.933	17.419	17.285	4.031	1.138	0.000
	0.472	0.372	0.350	0.684	0.643	0.707	0.802	0.779	0.773	0.180	0.051	0.000

TABLE XIX. Values of c' for histograms drawn from the standard deviation of the sizes of the known words.

	H	H1	H2	B	B1	B2	M	M1	M2	E	E1	E2
H	0.000	0.650	0.656	10.207	10.393	9.704	12.611	12.216	11.729	41.743	33.650	33.528
	0.000	0.009	0.009	0.080	0.083	0.078	0.103	0.106	0.101	0.478	0.479	0.478
H1	0.650	0.000	1.131	8.092	8.278	7.779	8.917	8.852	8.484	34.046	29.074	28.982
	0.009	0.000	0.017	0.089	0.092	0.086	0.100	0.102	0.098	0.470	0.470	0.469
H2	0.656	1.131	0.000	6.457	6.656	6.159	9.428	9.352	8.986	35.161	30.060	29.967
	0.009	0.017	0.000	0.071	0.074	0.069	0.107	0.109	0.104	0.487	0.488	0.487
B	10.207	8.092	6.457	0.000	6.831	6.683	29.218	22.346	21.408	65.138	46.483	46.284
	0.080	0.089	0.071	0.000	0.017	0.016	0.089	0.092	0.087	0.559	0.559	0.558
B1	10.393	8.278	6.656	6.831	0.000	11.703	30.425	24.194	23.312	64.556	46.384	46.188
	0.083	0.092	0.074	0.017	0.000	0.033	0.103	0.105	0.101	0.561	0.562	0.561
B2	9.704	7.779	6.159	6.683	11.703	0.000	22.666	18.113	17.199	63.969	45.943	45.748
	0.078	0.086	0.069	0.016	0.033	0.000	0.076	0.079	0.074	0.556	0.556	0.556
M	12.611	8.917	9.428	29.218	30.425	22.666	0.000	0.617	0.612	60.900	44.199	44.013
	0.103	0.100	0.107	0.089	0.103	0.076	0.000	0.003	0.003	0.541	0.541	0.540
M1	12.216	8.852	9.352	22.346	24.194	18.113	0.617	0.000	1.065	58.252	43.168	42.993
	0.106	0.102	0.109	0.092	0.105	0.079	0.003	0.000	0.006	0.541	0.542	0.541
M2	11.729	8.484	8.986	21.408	23.312	17.199	0.612	1.065	0.000	58.239	43.139	42.963
	0.101	0.098	0.104	0.087	0.101	0.074	0.003	0.006	0.000	0.540	0.541	0.540
E	41.743	34.046	35.161	65.138	64.556	63.969	60.900	58.252	58.239	0.000	0.250	0.251
	0.478	0.470	0.487	0.559	0.561	0.556	0.541	0.541	0.540	0.000	0.004	0.004
E1	33.650	29.074	30.060	46.483	46.384	45.943	44.199	43.168	43.139	0.250	0.000	0.434
	0.479	0.470	0.488	0.559	0.562	0.556	0.541	0.542	0.541	0.004	0.000	0.007
E2	33.528	28.982	29.967	46.284	46.188	45.748	44.013	42.993	42.963	0.251	0.434	0.000
	0.478	0.469	0.487	0.558	0.561	0.556	0.540	0.541	0.540	0.004	0.007	0.000

TABLE XX. Values of c' for histograms drawn from the sizes of the known words.

	H	H1	H2	B	B1	B2	M	M1	M2	E	E1	E2
H	0.000 0.000	3.011 0.135	2.550 0.114	11.538 0.516	10.800 0.483	9.816 0.439	7.692 0.344	6.373 0.285	6.082 0.272	21.509 0.962	21.551 0.964	21.525 0.963
H1	3.011 0.135	0.000 0.000	0.882 0.039	11.888 0.532	11.567 0.517	11.030 0.493	9.136 0.409	8.219 0.368	8.398 0.376	19.255 0.861	19.297 0.863	19.270 0.862
H2	2.550 0.114	0.882 0.039	0.000 0.000	11.261 0.504	10.624 0.475	10.132 0.453	8.343 0.373	7.426 0.332	7.583 0.339	19.700 0.881	19.741 0.883	19.715 0.882
B	11.538 0.516	11.888 0.532	11.261 0.504	0.000 0.000	1.632 0.073	1.878 0.084	5.724 0.256	6.283 0.281	7.133 0.319	22.338 0.999	22.361 1.000	22.338 0.999
B1	10.800 0.483	11.567 0.517	10.624 0.475	1.632 0.073	0.000 0.000	1.140 0.051	4.964 0.222	5.613 0.251	6.507 0.291	22.338 0.999	22.361 1.000	22.334 0.999
B2	9.816 0.439	11.030 0.493	10.132 0.453	1.878 0.084	1.140 0.051	0.000 0.000	4.092 0.183	4.629 0.207	5.434 0.243	22.338 0.999	22.361 1.000	22.334 0.999
M	7.692 0.344	9.136 0.409	8.343 0.373	5.724 0.256	4.964 0.222	4.092 0.183	0.000 0.000	1.722 0.077	1.901 0.085	22.315 0.998	22.338 0.999	22.312 0.998
M1	6.373 0.285	8.219 0.368	7.426 0.332	6.283 0.281	5.613 0.251	4.629 0.207	1.722 0.077	0.000 0.000	1.207 0.054	22.292 0.997	22.334 0.999	22.307 0.998
M2	6.082 0.272	8.398 0.376	7.583 0.339	7.133 0.319	6.507 0.291	5.434 0.243	1.901 0.085	1.207 0.054	0.000 0.000	22.315 0.998	22.338 0.999	22.312 0.998
E	21.509 0.962	19.255 0.861	19.700 0.881	22.338 0.999	22.338 0.999	22.338 0.999	22.315 0.998	22.292 0.997	22.315 0.998	0.000 0.000	2.472 0.111	3.751 0.168
E1	21.551 0.964	19.297 0.863	19.741 0.883	22.361 1.000	22.361 1.000	22.361 1.000	22.338 0.999	22.334 0.999	22.338 0.999	2.472 0.111	0.000 0.000	1.465 0.066
E2	21.525 0.963	19.270 0.862	19.715 0.882	22.338 0.999	22.334 0.999	22.334 0.999	22.312 0.998	22.307 0.998	22.312 0.998	3.751 0.168	1.465 0.066	0.000 0.000

TABLE XXI. Values of c' for histograms drawn from the mean of the sizes of the stopwords.

	H	H1	H2	B	B1	B2	M	M1	M2	E	E1	E2
H	0.000 0.000	4.327 0.194	5.241 0.234	7.849 0.351	6.015 0.269	7.558 0.338	8.832 0.395	6.909 0.309	6.529 0.292	20.951 0.937	20.970 0.938	20.917 0.935
H1	4.327 0.194	0.000 0.000	1.026 0.046	11.322 0.506	9.668 0.432	11.188 0.500	11.859 0.530	10.383 0.464	10.227 0.457	16.623 0.743	16.642 0.744	16.589 0.742
H2	5.241 0.234	1.026 0.046	0.000 0.000	11.663 0.522	10.030 0.449	11.596 0.519	12.177 0.545	10.724 0.480	10.567 0.473	15.710 0.703	15.729 0.703	15.676 0.701
B	7.849 0.351	11.322 0.506	11.663 0.522	0.000 0.000	2.974 0.133	1.476 0.066	2.996 0.134	2.929 0.131	2.862 0.128	22.315 0.998	22.334 0.999	22.281 0.996
B1	6.015 0.269	9.668 0.432	10.030 0.449	2.974 0.133	0.000 0.000	2.504 0.112	4.942 0.221	3.130 0.140	2.795 0.125	22.315 0.998	22.334 0.999	22.281 0.996
B2	7.558 0.338	11.188 0.500	11.596 0.519	1.476 0.066	2.504 0.112	0.000 0.000	2.639 0.118	1.789 0.080	1.655 0.074	22.338 0.999	22.334 0.999	22.281 0.996
M	8.832 0.395	11.859 0.530	12.177 0.545	2.996 0.134	4.942 0.221	2.639 0.118	0.000 0.000	2.281 0.102	2.661 0.119	22.338 0.999	22.334 0.999	22.281 0.996
M1	6.909 0.309	10.383 0.464	10.724 0.480	2.929 0.131	3.130 0.140	1.789 0.080	2.281 0.102	0.000 0.000	0.738 0.033	22.315 0.998	22.334 0.999	22.281 0.996
M2	6.529 0.292	10.227 0.457	10.567 0.473	2.862 0.128	2.795 0.125	1.655 0.074	2.661 0.119	0.738 0.033	0.000 0.000	22.315 0.998	22.334 0.999	22.281 0.996
E	20.951 0.937	16.623 0.743	15.710 0.703	22.315 0.998	22.315 0.998	22.338 0.999	22.338 0.999	22.315 0.998	22.315 0.998	0.000 0.000	4.870 0.218	6.237 0.279
E1	20.970 0.938	16.642 0.744	15.729 0.703	22.334 0.999	22.334 0.999	22.334 0.999	22.334 0.999	22.334 0.999	22.334 0.999	4.870 0.218	0.000 0.000	1.497 0.067
E2	20.917 0.935	16.589 0.742	15.676 0.701	22.281 0.996	22.281 0.996	22.281 0.996	22.281 0.996	22.281 0.996	22.281 0.996	6.237 0.279	1.497 0.067	0.000 0.000

TABLE XXII. Values of c' for histograms drawn from the standard deviation of the sizes of the stopwords.

	H	H1	H2	B	B1	B2	M	M1	M2	E	E1	E2
H	0.000 0.000	0.835 0.014	0.847 0.015	10.183 0.104	10.950 0.114	9.075 0.094	4.219 0.045	4.131 0.047	3.858 0.044	28.322 0.555	21.836 0.556	21.139 0.554
H1	0.835 0.014	0.000 0.000	1.456 0.029	8.314 0.119	8.916 0.128	7.563 0.109	4.081 0.060	4.064 0.061	3.857 0.058	24.599 0.540	19.810 0.541	19.248 0.539
H2	0.847 0.015	1.456 0.029	0.000 0.000	6.196 0.089	6.810 0.099	5.472 0.079	2.055 0.030	2.100 0.032	1.893 0.029	25.815 0.570	20.821 0.570	20.237 0.569
B	10.183 0.104	8.314 0.119	6.196 0.089	0.000 0.000	3.777 0.012	3.811 0.012	14.417 0.059	10.427 0.057	11.101 0.061	38.938 0.659	28.127 0.660	27.106 0.658
B1	10.950 0.114	8.916 0.128	6.810 0.099	3.777 0.012	0.000 0.000	6.571 0.024	15.289 0.069	11.548 0.067	12.190 0.070	39.275 0.669	28.450 0.669	27.424 0.668
B2	9.075 0.094	7.563 0.109	5.472 0.079	3.811 0.012	6.571 0.024	0.000 0.000	10.935 0.049	8.189 0.048	8.804 0.051	38.128 0.649	27.622 0.650	26.624 0.648
M	4.219 0.045	4.081 0.060	2.055 0.030	14.417 0.059	15.289 0.069	10.935 0.049	0.000 0.000	0.422 0.003	0.416 0.003	34.862 0.600	25.388 0.601	24.474 0.599
M1	4.131 0.047	4.064 0.061	2.100 0.032	10.427 0.057	11.548 0.067	8.189 0.048	0.422 0.003	0.000 0.000	0.725 0.005	34.184 0.602	25.154 0.602	24.267 0.601
M2	3.858 0.044	3.857 0.058	1.893 0.029	11.101 0.061	12.190 0.070	8.804 0.051	0.416 0.003	0.725 0.005	0.000 0.000	34.032 0.598	25.032 0.599	24.149 0.597
E	28.322 0.555	24.599 0.540	25.815 0.570	38.938 0.659	39.275 0.669	38.128 0.649	34.862 0.600	34.184 0.602	34.032 0.598	0.000 0.000	0.382 0.011	0.410 0.012
E1	21.836 0.556	19.810 0.541	20.821 0.570	28.127 0.660	28.450 0.669	27.622 0.650	25.388 0.601	25.154 0.602	25.032 0.599	0.382 0.011	0.000 0.000	0.686 0.023
E2	21.139 0.554	19.248 0.539	20.237 0.569	27.106 0.658	27.424 0.668	26.624 0.648	24.474 0.599	24.267 0.601	24.149 0.597	0.410 0.012	0.686 0.023	0.000 0.000

TABLE XXIII. Values of c' for histograms drawn from sizes of the stopwords.

B. Audio

This section presents c values drawn from audio for testing the sound system of the computer. The PCM samples of the files were normalized to fit the interval $[-1, 1]$ to yield the samples labeled. The wavelet decomposition was performed with the Daubechies 8 Wavelet function. The resulting values of the c statistic reflect most of all the different types of signals analysed: PCM samples and wavelet decomposition coefficients in different leaves. Among each type of signal, the type of sound is also reflected in the measures of c , with the noise having the highest values.

label	description	events
S1	recorded 'front center'	68545
W1-1	first wavelet approximation	31
W2-1	higher wavelet leaf	17147
S2	recorded 'front left'	71042
W1-2	first wavelet approximation	32
W2-2	higher wavelet leaf	17771
S3	recorded 'rear center'	65026
W1-3	first wavelet approximation	30
W2-3	higher wavelet leaf	16267
S4	recorded 'rear left'	63010
W1-4	first wavelet approximation	30
W2-4	higher wavelet leaf	15763
S5	noise	67579
W1-5	first wavelet approximation	31
W2-5	higher wavelet leaf	16906

TABLE XXIV. General description of the audio data used for the c values of the next table. The recorded data events are the PCM samples normalized to fit $[-1, 1]$. The wavelet first approximation consists of the low frequencies. The higher leaf consists of an approximation of one of the last details.

	S1	W1-1	W2-1	S2	W1-2	W2-2	S3	W1-3	W2-3	S4	W1-4	W2-4	S5	W1-5	W2-5
S1	0.000 0.000	1.788 0.321	22.671 0.194	8.340 0.045	2.062 0.365	27.502 0.232	15.078 0.083	2.054 0.375	24.858 0.217	11.170 0.062	2.108 0.385	28.275 0.250	36.639 0.199	1.893 0.340	19.829 0.170
W1-1	1.788 0.321	0.000 0.000	2.371 0.426	0.787 0.141	0.740 0.186	2.970 0.534	1.593 0.286	0.579 0.148	2.657 0.478	1.221 0.219	2.087 0.534	3.085 0.555	1.559 0.280	1.778 0.452	1.362 0.245
W2-1	22.671 0.194	2.371 0.426	0.000 0.000	17.933 0.153	2.289 0.405	3.937 0.042	31.633 0.272	2.456 0.449	2.242 0.025	26.022 0.224	3.180 0.581	5.087 0.056	42.198 0.361	2.399 0.431	29.940 0.325
S2	8.340 0.045	0.787 0.141	17.933 0.153	0.000 0.000	0.596 0.105	30.846 0.259	23.292 0.126	0.791 0.144	19.736 0.172	16.899 0.092	2.369 0.433	28.662 0.252	41.117 0.221	1.819 0.327	20.703 0.177
W1-2	2.062 0.365	0.740 0.186	2.289 0.405	0.596 0.105	0.000 0.000	3.030 0.536	1.499 0.265	0.730 0.185	2.944 0.521	0.702 0.124	1.664 0.423	3.142 0.556	1.030 0.182	1.540 0.388	1.087 0.192
W2-2	27.502 0.232	2.970 0.534	3.937 0.042	30.846 0.259	3.030 0.536	0.000 0.000	36.667 0.310	3.046 0.557	6.058 0.066	30.970 0.263	3.404 0.622	2.204 0.024	47.574 0.401	2.512 0.452	33.367 0.358
S3	15.078 0.083	1.593 0.286	31.633 0.272	23.292 0.126	1.499 0.265	36.667 0.310	0.000 0.000	1.691 0.309	32.883 0.288	12.918 0.072	1.711 0.312	36.572 0.325	27.398 0.151	1.667 0.299	24.678 0.213
W1-3	2.054 0.375	0.579 0.148	2.456 0.449	0.791 0.144	0.730 0.185	3.046 0.557	1.691 0.309	0.000 0.000	3.090 0.565	1.236 0.226	2.066 0.533	3.290 0.601	1.330 0.243	1.763 0.452	1.276 0.233
W2-3	24.858 0.217	2.657 0.478	2.242 0.025	19.736 0.172	2.944 0.521	6.058 0.066	32.883 0.288	3.090 0.565	0.000 0.000	28.223 0.248	3.218 0.588	4.879 0.055	44.388 0.388	2.425 0.436	31.727 0.348
S4	11.170 0.062	1.221 0.219	26.022 0.224	16.899 0.092	0.702 0.124	30.970 0.263	12.918 0.072	1.236 0.226	28.223 0.248	0.000 0.000	1.953 0.357	31.182 0.278	36.544 0.202	1.734 0.311	22.988 0.199
W1-4	2.108 0.385	2.087 0.534	3.180 0.581	2.369 0.433	1.664 0.423	3.404 0.622	1.711 0.312	2.066 0.533	3.218 0.588	1.953 0.357	0.000 0.000	3.431 0.627	2.272 0.415	1.377 0.353	2.746 0.502
W2-4	28.275 0.250	3.085 0.555	5.087 0.056	28.662 0.252	3.142 0.556	2.204 0.024	36.572 0.325	3.290 0.601	4.879 0.055	31.182 0.278	3.431 0.627	0.000 0.000	47.707 0.422	2.512 0.452	34.399 0.381
S5	36.639 0.199	1.559 0.280	42.198 0.361	41.117 0.221	1.030 0.182	47.574 0.401	27.398 0.151	1.330 0.243	44.388 0.388	36.544 0.202	2.272 0.415	47.707 0.422	0.000 0.000	2.322 0.417	17.089 0.147
W1-5	1.893 0.340	1.778 0.452	2.399 0.431	1.819 0.327	1.540 0.388	2.512 0.452	1.667 0.299	1.763 0.452	2.425 0.436	1.734 0.311	1.377 0.353	2.512 0.452	2.322 0.417	0.000 0.000	2.503 0.450
W2-5	19.829 0.170	1.362 0.245	29.940 0.325	20.703 0.177	1.087 0.192	33.367 0.358	24.678 0.213	1.276 0.233	31.727 0.348	22.988 0.199	2.746 0.502	34.399 0.381	17.089 0.147	2.503 0.450	0.000 0.000

TABLE XXV. Values of c for histograms drawn from sound PCM samples and wavelet leaf coefficients. The different types of the signals yield greater c values.

C. Music

This section presents measures of the c statistic drawn from the pitches of the notes of classical compositions. The results reflect music history. For example, measures of c involving Palestrina increases with the exception of Beethoven who, indeed, used modalism. The values of c related to Bach also increases along time, and the outcome of the comparison against Palestrina is only exceeded when Schönberg is reached, which reflects the non-tonal discourse of both Palestrina and Schönberg.

label	description	events
Pale	Sanctus 69 from G. P. da Palestrina	719
Bach1	BWV735 from J. S. Bach	236
Bach2	BWV648 from J. S. Bach	272
Moza1	K80 from W. A. Mozart	538
Moza2	K458 from W. A. Mozart	4218
Beet1	Opus 18, n1, mov. 3 from L. van Beethoven	1289
Beet2	Opus 132 from L. van Beethoven	17884
Schön	Opus 19, mov. 2 from A. Schönberg	102

TABLE XXVI. General description of the music data used for the c values of the next table. Each event is a midi value of a note pitch. Samples were chosen to reflect music history timeline. Works by the same composer were chosen among the first and last 10% of all he produced.

	Pale	Bach1	Bach2	Moza1	Moza2	Beet1	Beet2	Schön
Pale	0.00	1.36	1.39	2.05	2.44	1.72	2.33	1.89
Bach1	1.36	0.00	0.88	1.00	1.60	1.02	1.01	1.54
Bach2	1.39	0.88	0.00	1.73	1.22	1.41	1.68	1.52
Moza1	2.05	1.00	1.73	0.00	2.14	1.65	1.79	1.79
Moza2	2.44	1.60	1.22	2.14	0.00	1.62	4.20	1.84
Beet1	1.72	1.02	1.41	1.65	1.62	0.00	2.34	2.03
Beet2	2.33	1.01	1.68	1.79	4.20	2.34	0.00	2.02
Schön	1.89	1.54	1.52	1.79	1.84	2.03	2.02	0.00

TABLE XXVII. Values of c for histograms drawn from the pitches of classical compositions.

D. OS status

This last example expose the statistic c for samples drawn from the operational system of my laptop. The patterns are less neat than on last examples, but many conclusions can still be reached. The memory used by the most consuming processes compose the samples which present the highest values of c . Lowest values of c are related to RAM usage. Again, the type of samples are mandatory: they might all be identified by the values of c found in comparison to other samples, with the exception of the RAM memory.

label	description	events
cpu1	workload of the most active processor	888
cpu2	workload of the second most active processor	422
cpu3	workload of the third most active processor	1046
mem	RAM use in kB	557
p1	workload use of most consuming process	1197
m1	RAM use of most consuming process	1197
p2	workload use of second most consuming process	1197
m2	RAM use of second most consuming process	1197
p3	RAM use of third most consuming process	1197
m3	workload use of third most consuming process	1197

TABLE XXVIII. General description of the laptop system status data used for the c values of the next table. Each event is a measure in a snapshot of system status.

	cpu1	cpu2	cpu3	mem	p1	m1	p2	m2	p3	m3
cpu1	0.00	5.13	4.73	4.84	8.57	11.03	1.84	9.25	3.39	9.94
cpu2	5.13	0.00	2.97	3.83	8.89	9.90	4.72	9.03	3.64	9.00
cpu3	4.73	2.97	0.00	3.69	8.63	13.44	4.63	12.50	4.51	12.00
mem	4.84	3.83	3.69	0.00	4.16	10.77	4.60	9.53	3.97	9.48
p1	8.57	8.89	8.63	4.16	0.00	13.75	10.08	12.77	9.85	12.53
m1	11.03	9.90	13.44	10.77	13.75	0.00	12.26	12.53	10.57	15.43
p2	1.84	4.72	4.63	4.60	10.08	12.26	0.00	10.51	2.82	10.85
m2	9.25	9.03	12.50	9.53	12.77	12.53	10.51	0.00	10.30	8.67
p3	3.39	3.64	4.51	3.97	9.85	10.57	2.82	10.30	0.00	10.30
m3	9.94	9.00	12.00	9.48	12.53	15.43	10.85	8.67	10.30	0.00

TABLE XXIX. Values of c for histograms drawn from laptop system resource status measures.

IV. CONCLUSIONS AND FURTHER BENCHMARKS

The c statistic is robust both to determine if the distributions underlying the samples are the same and to quantify the difference between such probability distributions. The benchmarks for c , given in Section II, are very useful as references to make sense of data e.g. as the example analyses of Section III. Notice that the calculations require no training or clusterization usually involved in classification routines.

The rendering of this article and all tables are automated through Python scripts in order to ease the scrutinization of different settings³. After some exploration of the results, this setting was settled as template for this document: simulations used $N_c = 100$ comparisons per measure with $n = n' = 1000$ elements in each sample; all empirical density functions derived from histograms with $N_b = 30$ equally spaced bins. The main reason for this choice is that the results are consistent and stable, but are still of very modest scales. The use of more bins should enhance the results with respect to generality. On the other hand, samples are often not so big, which favors reporting the tables with a small sample size.

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