EVALUATION OF NUMERICAL EVIDENCE IN FORENSIC COMPARISON ANALYSES CRITERIA FOR IDENTITY AND ASSESSMENT OF EVIDENTIAL VALUE

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ABSTRACT

Two alternative methods of evaluating a discrepancy index are discussed, with a view to match crime/clue material with a comparison/source sample. One is put as $\theta = \sum_{i=1}^{N} (Q_i - 1)^2/q_i^2$ (cf. Parker's C) and another, $\Gamma = \prod_{i=1}^{N} Q_i$ (1 - $t_i q_i$) based on the discrepancy quotient Q and the overall coefficient of variation q. A simple evaluation of the uniqueness of a crime material is described on the basis of attributes measured in the source and the population, their mean values and the respective standard deviations.

To determine if a "crime sample" has originated from an alleged source. The inference drawn from non-numerical analyses like tool marks examination, e.g., in forensic ballistics or hand-in-glove fit of a crime sample with its remnant available elsewhere, can (when good characteristic matches are observed) be interpreted directly without "any reasonable doubt". This however is not possible in (numerical) material analyses. In the former the characteristic match is obvious, but in the latter one has to cautiously interpret the results. Forensic analytical data have been evaluated by Parker² (human hair), Hoffman³ (soils) and Lawson and Framan⁴ (general).

Evaluation of analytical data in forensic problems is considered here firstly to establish if the crime and comparison (alleged source) samples match and secondly, that if these do so, then what proportion of the population of such material in question attain the levels of the attributes analysed. The lower it is, the more weighty is the evidence.

1. CRITERIA FOR IDENTITY

For matching or otherwise of a crime sample with the source, Parker² defined the discrepancy index, C, as:

$$C = \sum_{i=1}^{N} D_i^2 = \sum_{i=1}^{N} (X_i - Y_i)^2 / \lambda_i^2$$
 (1)

where X_i and Y_i are the values for the *i*-th attribute measured in the crime sample and the alleged source respectively. λ_i is the combined standard deviation resulting from the standard deviation δ_1 in the crime sample, due to the measurement errors and the standard deviation δ_2 in the source, mainly due to the intrinsic variability; $\lambda^2 = \delta_1^2 + \delta_2^2$. If N uncorrelated attributes are measured, C is distributed as χ^2 with N degrees of freedom, if the crime sample originates from the alleged source. The observed C may be compared with the critical value for C, viz., C_0 , from the statistical χ^2 tables. At a given chosen probability value, say 0.01 (Parker),

if $C < C_0$ then the crime sample could stem from the alleged source, and if in fact it does, the probability of finding $C > C_0$ is as low as 0.01.

An alternative to Parker's C is described here, based on discrepancy quotient (vide infra) which is significant to the forensic analyst. Moreover, Parker used $\ln c$ as against c (concentration); this is laborious, without making any significant difference in the results using c conveniently.

Let the values of a given attribute in the crime sample and the alleged source be: $X_1 \pm \delta_1$ and $X_2 \pm \delta_2$ respectively, and let $\delta_1/X_1 = q_1$ and $\delta_2/X_2 = q_2$. We define discrepancy quotient, Q, as $Q = X_1/X_2$, we have $(X_1 \pm \delta_1)/(X_2 \pm \delta_2)$ $= Q \pm k$, where $k/Q = \sqrt{q_1^2 + q_2^2} = q$ (let). Parker's element of discrepancy index, viz., $D_i^2 =$ $(X_1 - X_2)^2 / (\delta_1^2 + \delta_2^2)$ can now be put as : $D_i^2 =$ $(Q-1)^2/(Q^2q_1^2+q_2^2)$. Discrepancy quotient, Q, will be useful in our discussion, if we impose the condition that $Q \ge 1$, no matter whether the attribute in the crime sample is greater than that in the source or vice versa. It will indicate a Qfold discrepancy in the value of the attribute in the crime sample and source and calculations of the discrepancy index (vide infra) will be to the benefit of the accused. We now define \triangle_i , a normal deviate with unit standard deviation as

 $\Delta_i^2 = (1/Q - 1)^2/(k')^2$, where k'/(1/Q) = q. •• if $(X_2 \pm \delta_2)/(X_1 \pm \delta_1) = (1/Q \pm k')$, we have, k'/(1/Q) = q. For identity of crime sample and source, we have $Q \rightarrow 1$, $(1/Q - 1) \rightarrow 0$. From the above we get

$$\Delta_i^2 = (Q_i - 1)^2/q_i^2 \tag{2}$$

The modified discrepancy index, θ , may now be compared with the critical value θ_0 from χ^2 tables to test the match.

$$\theta = \sum_{i=1}^{N} \Delta_i^2 = \sum_{i=1}^{N} (Q_i - 1)^2 / q_i^2. \tag{3}$$

when

$$Q \rightarrow 1, \ \wedge_i{}^{i} \rightarrow D_i{}^{i}$$

It may appear that when X_1 and X_2 are (presumed) to be normally distributed. D 2 is χ^2 distributed. then Δz^2 cannot be χ^2 distributed. In fact "a Gaussian distribution may be inappropriate for measurements, such as chemical composition, which must be always non-negative, a functional transform whereby the new variate is brought into closer alignment with the Gaussian distribution may be used instead. The logarithm of the chemical composition could be used rather than the chemical composition itself^{2(a)}. Indeed the logconcentration of elements in human hair has been found to be closely Gaussian^{2(b)}. On the first presumption $X_2 - X_1$ is a Gaussian variate with zero mean and standard deviation $\sqrt{\delta_1^2 + \delta_2^2}$. Now, if on the other hand $\ln X_1$ and $\ln X_2$ are admittedly better Gaussian distributed, and if the sample could stem from the alleged source,

$$\ln X_2 - \ln X_1 = \ln (X_2/X_1) = \ln \left(1 + \frac{X_2 - X_1}{X_1}\right) \approx \frac{X_2 - X_1}{X_1} = \left(\frac{1}{\bar{Q}} - 1\right)$$

is a Gaussian variate with zero mean and standard deviation k' = q/Q, and Δ_i^2 is χ^2 distributed. The approximation involved is valid at lower discrepancies where the test for matching is sought.

It will now be shown that the use of c in place of $\ln c$ (cf. Parker) does not make any significant difference where a critical examination is necessary. For large discrepancies, the difference in the crime sample and the alleged source is obvious; an inspection of data is sufficient. The use of c units makes evaluation of data less laborious.

Using c units, we have a series of measurements:

 $c_1 = \overline{c} + x_1, \ldots, c_n = \overline{c} + x_n$, where x's can be + ve - ve, yielding the result : $\overline{c} \pm \delta_c$ where δ_c is the standard deviation, calculated using c units directly. To convert the measurements into $\ln c$ units, we

have, $\ln c_1 = \ln (c + x_1) \approx \ln c + x_1/c$, etc., showing that $\delta_{\ln c} \approx \delta_c/c$ where $\delta_{\ln c}$ is the standard deviation calculated using $\ln c$ units. Parker's discrepancy index element, D_i^2 using $\ln c$ units, can now be readily shown $\approx D_i^2$ using c units, at lower discrepancies.

 $D_t^2 = (\ln X_1 - \ln X_2)^2/[(\delta_{c,t}/X_1)^2 + (\delta_{c,2}/X_2)^2]$ can be shown, at lower discrepancies, $\approx (X_1 - X_2)^2/(\delta_{c,1}^2 + \delta_{c,2}^2)$. Thus in Parker's eqn. (1) whether $\ln c$ values or c values are taken, these reduce to the same form and magnitude at lower discrepancies, where the test is necessary.

Alternative Method for Constructing a Discrepancy Index.—The discrepancy quotient, $Q, (Q \ge 1)$ may be suitably modified to yield a minimum value, $Q_{\min,i}$. Thus at a given chosen probability level

 $Q_{min.} = Q - t$. k. where 7—times the standard deviation is subtracted on the basis of Student's t value for the given desired probability level and for the given degrees of freedom. Thus for a series of 10 measurements (9 degrees of freedom) for i-th attribute the probability of a chance deviation in Q not exceeding $2.26 \ k$ is 95% (t = 2.26). Having obtained $Q_{min.}$ values for each attribute. ($Q_{min} \ge 1$), we have the alternative discrepancy index, which may now be denoted by Γ , simply given by:

$$\Gamma = \prod_{i=1}^{N} Q_{min, i} = \prod_{i=1}^{N} (Q_i - t_i k_i)$$

$$= \prod_{i=1}^{N} Q_i (1 - t_i \cdot q_i). \tag{4}$$

 $Q_{min.}$ values $\geqslant 1$ only are considered; those working out to < 1 are ignored, as these mean that there

is no significant difference in the values of the given attribute. Ideally, Γ_0 for identity between the crime sample and the alleged source should be unity. Values of Γ in excess of unity would therefore indicate the extent of discrepancy between the two.

Application of the above criteria to typical literature data is given in Tables I to III. While Γ is well suited to the evaluation of data on chemical composition it will be apparent that in the evaluation of data on physical properties like density, refractive index, etc., in the event of high discrepancies (owing to very low q's), Γ will not suitably reflect these. The discrepancy indexes c and θ have therefore general applicability.

2. Assessment of Evidential Value

Once the questioned material is shown indistinguishable from the alleged source, the next consideration arises as to what significance or evidential value this agreement has. Let p_i be the probability of finding the level of the *i*-th attribute in the entire population similar to that in the alleged source then the probability of finding N uncorrelated attributes (which may be measured) in levels

similar to those in the alleged source is $P_N = \prod_{i=1}^{N} p_i$. The smaller the value of P_N , the more weighty is the evidence. The value of an attribute towards identification depends on "the ratio of the measurement error to the spread of the attribute over the population and also on the value of the sample measurement when referred to the peak of the frequency distribution" $2^{(b)}$. In the case of normal distributions the probability p_{ij} of finding the level of the attribute i in the entire population similar

TABLE I

Sample 1 w.r.t. "control" (NAA of human hair) 2 (b)

		$D_{\mathbf{f}}^{2}$				$Q_{min} \sim 1$	
Attribute	(estimated)	Using In c units	Using c units	^ ²	Q >1	= Q (1-1.38 q)	$= \frac{(P = 95\%)}{Q(1-2\cdot26 q)}$
Na	0.24	4 · 1	2 · 5	6.2	1.60	1.07	†
Zn	0.10		se for Cl. N 2·9	1n, I, Ču, Br, A 2·9		1.01	~~
N=9		$\Sigma = 15 \cdot 1$	$\Sigma = 12.4$	$\mathcal{E} = 22 \cdot 8$	<u></u> -	$\Pi = 1 \cdot 22$	Π == 1·0

[†] a dash denotes $Q_{min} = 1.00$.

Table II

Summary of analysis of data on samples Nos. 1, 4, 15, 30 w.r.t. "control" (NAA of human hair)^{2(b)}

Sample No.	No. of attributes measured		N Σ D ² In c units	$\overset{\mathtt{N}}{\mathcal{\Sigma}}$ \wedge	I_i^2	ŽQ _i (1−2·2	Remarks 26 _{qi})
1	9	15.1	(12.4)†	22.8	(12.3)*	1.0	Source similar to control
4	10	42.9	(33 · 9)†	111.0	(34.0)*	1 · 28	Source different, significant dis- crepancy
15	10	209-2		2760-		17-2	Source different, large discrepancy
30	9	2.613	> 10 ³	2·30 _s	103	108 · 3	Source different, v. large discrepancy

[†] recalculated values, using c units, instead of In c units.

Table III

Soils (F and K-1)— Evaluation of spectrographic (Seidel function) data⁵

Attribute	$q_{\mathbf{t}}$	q_2	ų	D ²	△;	Q>1	$= \frac{Q_{\min} 1.1}{Q(1-2\cdot 26_q)}$
Mn	0.20	0-21	0.29	11.6	92.9	3·79 ₅	1 · 31
		Līkewi	ise for Al, Si,	Cr, Na, Fe	, Mg, Cu, C	a	
Ti	0.25	0.21	0.32;	15.5	259 · 5	6.26	1 · 63 ₅
N=10			Σ	$rac{1}{2}$	$\Sigma = 629 \cdot 5$		$\frac{10}{11} = 4 \cdot 10$

Conclusion: Soils from different sources.

to that in the crime material or the alleged source, for which the intrinsic variability of the material and the measurement errors are determined would simply be given by the ratio of areas under the respective distribution curves, a/A, where a is that under the source curve and A is that under the population curve. If the latter is plotted as ϕ_x is

or curve (Fig. 1) such that the normal distribution is valid: $\phi_a = (1/\sqrt{2\pi}) \exp(-x^2/2)$ where x is measured in standard deviation σ units, we have, by definition, area under population curve (PQR, Fig. 1), A = 1. Therefore $p_i = a$, the area under the distribution curve for the source (say, ABC or DEF) which is properly constructed within the

 $[\]Sigma_0 = 27.8$, $\Sigma_0 = 29.6$ (at 0.1% probability level for $C > C_0$ or $\theta > \theta_0$)

^{*} Calcd. without imposing the condition $Q \ge 1$; $Q = (X_1/X_2) \ge 1$.

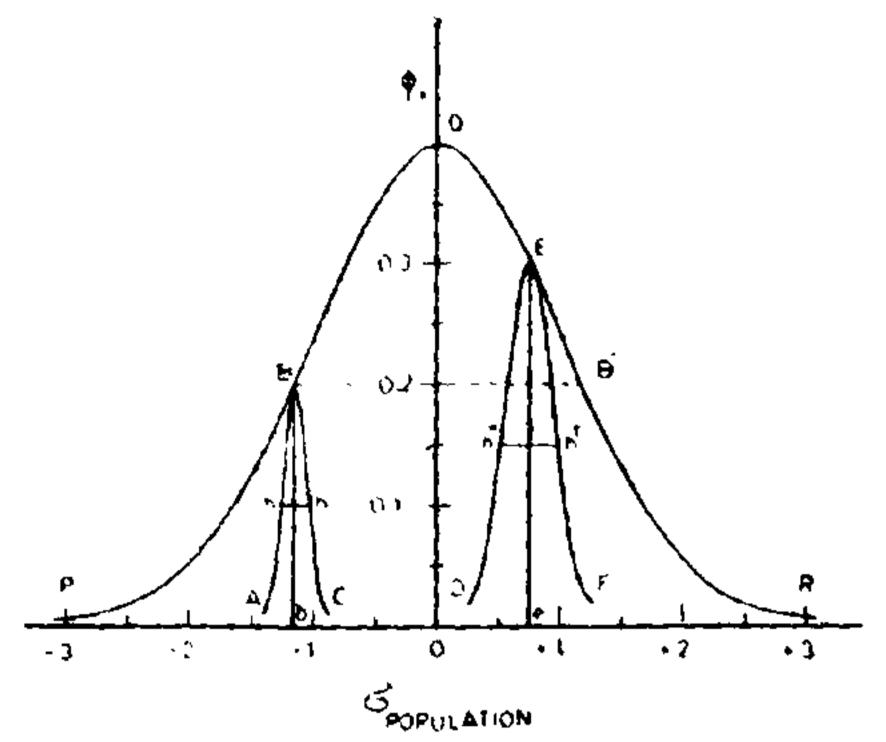


Fig. 1. Construction of a typical normal distribution curve for the alleged source material (ABC or DEF) within that of the population (PQR), p_{ij} t in $\sigma_{population}$ units, $\sigma_{source}/\sigma_{population} = 0.04_7$, -1.17, 0.1 and 0.14, +0.75, 0.2 for curves ABC and DEF respectively.

The approximation in the above equation may be corrected for by utilizing:

$$p_i = 2.51 \, \phi^*_{x, i} \, (\sigma_{\text{source}} / \sigma_{\text{population}})_i \tag{5}$$

with benefit to the accused. ϕ_{xo} (ϕ corresponding to $\sigma = 0$) = 0.3989, is equivalent to area, A = 1 (vide supra). Therefore, area equivalent to $\phi *_x = (1/0.3989) \phi *_x = (\sigma_{source}/\sigma_{population})$, the same as in eqn. (5).

Illustration.—The calculation of P_N for a sample human hair^{2(b)} is illustrated in Table IV. The analysis utilizing eqn. (5) would thus reveal, granted that our assumption of $(\log -)$ normal distribution over the population is valid and that the attributes are uncorrelated, that there is 1 in $\sim 10^7$ chance of finding such hair in the population in question, with levels of the ten attributes in question similar to those in the given sample. The magnitude of P_N is indicative of the uniqueness of the given physical evidence material. It is an attempt to quantitate the evidential value.

Table IV

Uniqueness of crime material—Human—hair in "control" sample2(b)

Attribute	^μ control - ^μ population	ϕ^*	$p_i = 2.36 \ \phi^*_x \left(\frac{\sigma_{\text{control}}}{\sigma_{\text{population}}} \right)$	
	σpopulation = t	ϕ^* (corresponding to t)		
Na	0.442	0.362	0 · 201	
Sb	0.855	Cl, Mn, I, Cu. Br, Au. Hg. Zn 0.277	0.284	

population curve. Evaluate $t = |\mu_{population} - \mu_{source}|/\sigma_{population}$, where μ is the mean value. The crime material or the source is therefore t units of population-standard deviations removed from the population mean value. Construct a normal curve on the ordinate at t converting σ_{source} into $\sigma_{population}$ units. by the factor $(\sigma_{source}/\sigma_{population})$. The area under the source curve, a, can be evaluated as $a = \phi_{\sigma}$ (H.I.B.W.) where ϕ_{σ} is the ordinate corresponding to the t value (e.g., Bb or Ee) which can be directly read from the normal error curve tables. The half-intensity band width, H.I.B.W., $(e.g., hh' \text{ or } h^*h^*) = 2.36 \ (\sigma_{source}/\sigma_{population})$ for the source curves. H.I.B.W. $= 2.36 \ \sigma$ units for any normal curve. We have, therefore,

$$p_i = a_i = 2.36 \, \phi_{x,i} \, (\sigma_{\text{source}}/\sigma_{\text{population}})_i$$

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