A NOTE ON THE DISTRIBUTION OF DIFFERENCES IN MEAN VALUES OF TWO SAMPLES DRAWN FROM TWO MULTIVARIATE NORMALLY DISTRIBUTED POPULATIONS, AND THE DEFINITION OF THE D*-STATISTIC.

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1. I am indebted to Professor S. N. Bose of Dacca for pointing out a mistake which occurred in my paper "On the Exact Distribution and Moment-coefficients of the Destatistic" published in Volume 2, Part 2 of this journal, in writing down the distribution of the set of mean differences of two samples drawn from two multi-variate normally distributed populations. I obtain in this note the correct distribution which necessitates a small alteration in the definition of the D*-statistic for the case of correlated variates. The net result is that the formulae (1·6), (1·7), (2·2) and (3·1) of my previous paper need modifications, which I give here; but the main investigation remains completely valid.

It is also shown that if we assume the two populations to have the same set of variances and covariances, no modifications in the results given previously are necessary. As in practical applications of the D²-statistic, it is usually assumed that the variances and covariances are identical in the two populations, the results obtained in the previous paper can be used legitimately.

 Let Σ and Σ' be two random samples of sizes n and n' drawn respectively from two normal populations n and n' of p-variates which are linearly correlated. We shall write down the population statistics in the following way:

$$a_i = \text{mean value of the } i\text{-th character in population } \pi$$

$$a'_i = \text{mean value of the } i\text{-th character in population } \pi'$$

$$a_i = \text{standard deviation of the } i\text{-th character in population } \pi'$$

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 ρ_0 = coefficient of correlation between the *i*-th and *j*-th character in population π' $\rho'_0 = \text{coefficient of correlation between the$ *i*-th and*j* $-th character in population <math>\pi'$ $\rho''_0 = \text{coefficient of correlation between the } \rho''_0 = \rho''_0$

We also write $\alpha_{ij} = \sigma_i \sigma_j \rho_{ij}$, $\alpha'_{ij} = \sigma_i' \sigma_j' \rho'_{ij}$... (2.4)

The D*-statistic was intended to be and was defined as a quantity determined entirely by the sample values of the variates. Raj Chandra Rose has investigated the exact distribution of a modified form of the D*-statistic in which the population values of the variances and co-variances have been substituted for the corresponding sample estimates.—Editor, Sankhyd.

For the two samples Z and Z' we write

Let Am be the co-factor of am in the determinant

and A'm the co-factor of a'm in the determinant .

$$A' \equiv \begin{bmatrix} \alpha'_{11} & \alpha'_{12} & \dots & \alpha'_{1p} \\ \alpha'_{21} & \alpha'_{22} & \dots & \alpha'_{2p} \\ \dots & \dots & \dots & \dots \end{bmatrix}$$
... (2.7)

Then the distribution of the set of mean differences (a, a,a) in repeated samples of size n drawn from the population w may be written as

$$-(n/2\Lambda)[\Lambda_{11}(a_1-\alpha_1)^2+...+2\Lambda_{12}(a_1-\alpha_1)(a_2-\alpha_2)+...]$$
Constant × e

$$...da_1.da_2...da_n... (2.8)$$

and likewise the distribution of the set of mean differences (a1', a2',a) in repeated samples of size n' drawn from the population n' may be written as

$$\begin{array}{ll} -(\pi/2\Lambda')\{\Lambda'_{11}(a_1'-a_1')^2+.....+2\Lambda_{13}'(a_1'-a_1')(a_2'-a_3')+.....\}\\ \text{Const.} \times \epsilon & ...da_1' da_2'.....da_s' \ (2^{\circ}0) \end{array}$$

Our immediate object is to write down the distribution of the set of mean differences $(a_1-a_1', a_2-a_2', \ldots, a_n-a_n').$

3. Let us set
$$(a_1-a_1)+(a_1'-a_1')=2x_1$$
 $(i=1,2,.....p)$... (3·1)
$$(a_1-a_1)-(a_1'-a_1')=2y_1$$
 $(i=1,2,.....p)$... (3·2)
$$(a_1'-a_1')=(x_1-y_1)$$
 $(i=1,2,.....p)$... (3·2)

We shall further set
$$r_{ij} = \frac{n\Lambda_{ij}}{2\Lambda}$$
, $r'_{ij} = \frac{n'\Lambda'_{ij}}{2\Lambda'}$... (3.3)

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Then from (1.7) and (1.8) the joint distribution of x_1, x_2, \dots, x_p ; y_1, y_2, \dots, y_n may be written as

Const.
$$\times \exp \{ -\gamma_{11}(x_1 + y_1)^2 - \dots - 2\gamma_{12}(x_1 + y_1) (x_2 + y_2) - \dots - \gamma_{11}'(x_1 - y_1)^2 - \dots - 2\gamma_{11}' (x_1 - y_1) (x_2 - y_2) - \dots \} dx_1 dx_2 \dots dx_p dy_1 dy_2 \dots dy_p (3.4)$$

or Const.
$$\times exp \left[- \left[(\gamma_{11} + \gamma_{11}')x_1^2 + \dots + 2(\gamma_{12} + \gamma_{12}')x_1x_2 + \dots \right] - 2\left[x_1(\gamma_{11}y_1 + \gamma_{12}y_2 + \dots + \gamma_{12}y_2 + \dots - x_1(\gamma_{11}y_1 + \gamma_{12}y_2 + \dots + \gamma_{12}y_2) + \dots \right] - \left[(\gamma_{11} + \gamma_{11}')y_1^2 + \dots + 2(\gamma_{12} + \gamma_{12}')y_1y_2 + \dots \right] \right] dx_1 dx_2 \dots dx_2 dy_1 dy_2 \dots dy_2$$
 (3.5)

4. The expression within the squared brackets in (3.5), when equated to zero, may be looked upon as representing a hyperquadric in a space of p-dimensions; x₁, x₂,...x_p being regarded as the variables, and y₁, y₂, ...y_p being momentarily regarded as constants.

Hence there exists a linear transformation by which the origin is transformed to the centre of this quadric. This transformation is of the type $x_1 = x_1' + \text{const} \dots$ (4.1)

Making this transformation, the joint distribution of x_1' , x_2' , ... x_p' ; y_1 , y_2 , y_n can be written

Const.
$$\times exp \left[-\{(\gamma_{11} + \gamma_{11}^{-1})x_{1}^{-1} + \dots + 2(\gamma_{11} + \gamma_{12}^{-1})x_{1}^{-1}x_{1}^{-1}x_{1}^{-1} + \dots + (\gamma_{11} + \gamma_{12}^{-1})x_{1}^{-1}x_{1}^{-1}x_{1}^{-1} + \dots + (\gamma_{11} + \gamma_{11}^{-1})x_{1}^{-1}x_{1}^{-1}x_{1}^{-1} + \dots + (\gamma_{11} + \gamma_{11}^{-1})x_{1}^{-1}x_{1}^{-1}x_{1}^{-1}x_{1}^{-1} + \dots + (\gamma_{11} + \gamma_{11}^{-1})x_{1}^{-1}x_{1}^{-$$

where
$$k \equiv \begin{bmatrix} \gamma_{11} + \gamma_{12}^{\prime} & \gamma_{12} + \gamma_{12}^{\prime} & . & \gamma_{19} + \gamma_{19}^{\prime} \\ \gamma_{21} + \gamma_{21}^{\prime} & \gamma_{22} + \gamma_{22}^{\prime} & . & \gamma_{29} + \gamma_{29}^{\prime} \\ ... & ... & ... \\ \gamma_{21} + \gamma_{21}^{\prime} & \gamma_{22} + \gamma_{22}^{\prime} & . & \gamma_{29} + \gamma_{29}^{\prime} \end{bmatrix} ... (4.3)$$

where
$$\psi \equiv \gamma_{11} y_1^0 + \dots 2\gamma_{12} y_1 y_2 + \dots = \sum_{i,j=1}^{n} \gamma_{ij} y_1 y_2 \dots$$
 (4.5)

$$\phi' \equiv \gamma_{11}' y_1^a + 2 \gamma_{12}' y_1 y_2 + = \sum_{i,j=1}^{9} \gamma_{ij} y_i' y_j ... (4.6)$$

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To simplify (4.4) we multiply, the 1st, 2nd, pth column of the determinant by y_1, y_2, \dots, y_p and add to the (p+1)th column, then from Euler's theorem

Again multiplying the 1st, 2nd, pth row by $y_1, y_2,, y_p$, and subtracting from the last row we get on again applying Euler's theorem

In (4.2) we can integrate out for x_1', x_2', \dots, x_s' . Hence we get as the distribution of y_1, y_2, \dots, y_s

Const.
$$\times e^{-\phi(y_1, y_2, ..., y_p)/k} dy_1 dy_2, ..., dy_p$$

5. If k_0 denotes the cofactor of the element in the *i*-th row and the *j*-th column of the determinant k given by (4.3) we have

$$\varphi(y_1, y_2, \dots y_p) = \sum_{i \neq j = 1}^{p} k_{ij} \frac{\partial \psi}{\partial y_i} \frac{\partial \psi}{\partial y_j} = 4 \sum_{i,j=1}^{p} l_{ij} y_i y_j \dots (5.1)$$

$$l_{mq} = l_{nm} = \sum_{i=1}^{n} k_{i1} \gamma_{mi} \gamma_{nj}^{i}$$
 ... (5.2)

Substituting for y_1, y_2, \dots, y_p , from (3.1), and using (4.9) and (5.1); we can write the distribution of $(a_1 - a_1), a_2 - a_2, \dots, a_p - a_p$ in the form

$$c \frac{(-1/k)[l_{11}\{(a_1-a_1')-(\alpha_1-\alpha_1')]^2+...+2l_{12}[(a_1-a_1')-(\alpha_1-\alpha_1')][(a_2-a_2')-(\alpha_2-\alpha_2')\}...]}{\times d(a_1-a_1')d(a_2-a_2')} \times d(a_1-a_1')d(a_2-a_2')(5:3)$$

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6. Hence the formula (2.2) of my paper "On the Exact Distribution and Moment Coefficients of the D'-statistic": will be valid provided that we set

$$\beta = k$$
, $\beta_{ij} = \frac{4l_{ij}}{\tilde{n}}$... (6·1)

where k is defined by the relation (4.3) of this note, In by the relation (5.2), and

$$\frac{2}{n} = \frac{1}{n} + \frac{1}{n'} \dots (6.2)$$

The formulæ (1.6), (1.7) on page 145, Sankhyā vol. 2, part 2 should be dropped. The definition (3.1), on page 146 for the Di-statistic remains valid, provided that by \$\beta\$ and β_u we understand the constants given by the relation (6.1) of the present note. No further corrections are necessary in the remainder of the paper.

7. Let us now go on to consider the special case

i.e. when the two populations w and w' have the same set of variances and covariances.

From (3.3) we see that;
$$n\gamma'_{ij} = n'\gamma_{ij}$$
 ... (7.2)

From (3'3) we see that;
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 ... (7'2) and hence from (4'5) and (4'6) $n'\psi' = n\psi'$... (7'3)

and also
$$n'\frac{\partial \psi}{\partial v_i} = n\frac{\partial \psi'}{\partial v_i}$$
 $(i = 1, 2, p)$... (7.4)

If we substitute for y'a and 34/3y, from (7.3) and (7.4) in (4.8); then multiply the 1st, 2nd, pth column of the determinant by

$$\frac{2n}{n+n_1}y_1, \frac{2n}{n+n_1}y_2, \dots, \frac{2n}{n+n_1}y_2 \dots \dots (7.41)$$

and subtract their sum from the last column, we see that

$$\phi(y_1, y_1, \dots, y_n) = \frac{4n!}{n+n!} \psi \cdot \left(\frac{n+n!}{n!}\right)^n \mid y_{ij} \mid \dots$$
 (7.42)

where

Again
$$K = \left(\frac{n+n'}{n}\right)^n | \gamma_{ij} |$$
 ... (7.44)

Hence (4.0) can be written as Const.
$$\times e^{-4n'\psi/(n+n')} dy_1 dy_2 \dots dy_p$$
 ... (7.45)

Hence from (4.5), (3.3) and (6.2), we can write the distribution of y_1, y_2, \dots, y_p in the form

$$\begin{array}{c} -(\bar{n}/\Lambda)\{\Lambda_{1},y_{1}^{2}+.....2\Lambda_{1},y_{1},y_{2}+.....\}\\ \text{Const.} \times \epsilon & ...dy_{1}, dy_{2},....dy_{2} &(7.5) \end{array}$$

Finally from (3*1), we see that the distribution of $(a_1-a_1', a_2-a_2',a_p-a_p')$ can be written in the following form

$$C.s = \frac{-(\bar{u}/4\Lambda)[\Lambda_{11}\{(a_1-a_1')-(a_1-\bar{u}_1')\}^2 + ... + 2\Lambda_{12}[(a_1-a_1')-(a_1-a_1')](a_2-a_2') - (a_1-a_2')] + 1}{\times d(a_1-a_1')d(a_2-a_2)....d(a_2-a_2)} (7.6)$$

In this special case therefore, the results of my previous paper remain valid, without

In this special case therefore, the results of my previous paper remain valid, without any correction; since in this case σ_{m} as defined on page 145, formula (1'6) of paper, is simply equal to σ_{m} , and consequently $\beta \equiv \Lambda$ and $\beta_{ij} \equiv \Lambda_{ij}$.

Thus if two populations π and π' have the same set of variances and covariances, then we can define D^2 by $D^2 = D_1^2 - 2/\bar{n}$... (7.7)

where
$$D_1^2 = \frac{1}{P\Lambda} \{ \Lambda_{11}(a_1 - a_1')^2 + \dots 2\Lambda_{12}(a_1 - a_1')(a_2 - a_2') + \dots \}$$
 ... (7.8)

8. If the variables in both the populations are independent. Then

$$\gamma_{ij} = \gamma_{ij}' = 0 \quad \text{when } i \neq j \; ; \; \gamma_{ii} = \frac{\pi}{2} \quad \frac{\Lambda_{ii}}{\Lambda} = \frac{\pi^2}{2\sigma_{ii}} \; ; \quad \gamma_{ii}' = \frac{\pi}{2} \quad \frac{\Lambda_{ii}'}{\Lambda'} = \frac{\pi}{2\sigma_{ii}'}$$

$$k_{ij}=0 \ \ \text{when} \ \ i\neq j, \ k_{ii}=\frac{k}{\gamma_{ii}+\gamma_{ii}}, \ \ l_{ii}=0 \ \ \text{when} \ i\neq j, \ l_{ii}=k_{ii}\gamma_{ii}'=\frac{k\gamma_{ii}\gamma_{ii}'}{\gamma_{ii}+\gamma_{ii}'}$$

Therefore, $\beta_{ij} = 0$ when $i \neq j$, $\beta_{ii} = \frac{4}{3} \cdot \frac{k r_{ij} r_{ij}'}{r_{ij} + r_{ij}'}$

or
$$\beta_{ii} = \frac{4k}{\tilde{n}} \cdot \frac{1}{(1/\gamma_{ii}) + (1/\gamma_{ii})} = \frac{n + n'}{n'\alpha_{ii} + n\alpha_{ii}}$$

Therefore,
$$D^2 = -\frac{1}{p\beta} \{\beta_{11}(a_1 - \dot{a}_1')^2 + \dots \}$$

= $-\frac{1}{p\beta} \{\frac{(a_1 - a_1')^2}{\frac{n}{n+n'}!} + \dots \}$

Thus for the uncorrelated case the generalised definition agrees with the one originally given by Mahalanobis.

May, 1938. Statistical Laboratory, Calcutta.