

ON A MEASURE OF DIVERGENCE BETWEEN TWO MULTINOMIAL POPULATIONS.

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INTRODUCTION

The problem of discrimination in Statistics is not new. It has repeatedly been stressed by Mahalanobis (1930) that mere "tests of significance" can not always answer all our questions and for this reason he introduced the idea of divergence of two populations. His "generalised distance" measures the divergence of two normal correlated multivariate populations with the same sets of variances and covariances. Bose and Roy (1938) found the distribution of D^2 , "Mahalanobis's generalised distance" in its studentised form.

2. DEFINITION OF DIVERGENCE

Let two multinomial populations be characterised by the two sets of probabilities (π_1, \dots, π_k) and (π'_1, \dots, π'_k) . Then, as $\sum \pi = 1$ and $\sum \pi' = 1$, $(\sqrt{\pi_1}, \dots, \sqrt{\pi_k})$ and $(\sqrt{\pi'_1}, \dots, \sqrt{\pi'_k})$ may be considered to be the direction cosines of two st. lines through the origin in a k -dimensional space. The square of the angle between these two lines may be considered to be an appropriate measure of divergence between the two multinomial populations. Thus, if the measure of divergence be denoted by Δ^2 then

$$\cos \Delta = \sqrt{\pi_1 \pi'_1} + \sqrt{\pi_2 \pi'_2} + \dots + \sqrt{\pi_k \pi'_k} \quad \dots (2.1)$$

This can be written as

$$4 \sin^2 \frac{\Delta}{2} = (\sqrt{\pi_1} - \sqrt{\pi'_1})^2 + (\sqrt{\pi_2} - \sqrt{\pi'_2})^2 + \dots + (\sqrt{\pi_k} - \sqrt{\pi'_k})^2 \quad \dots (2.2)$$

From this latter expression it is evident that when $\pi_i = \pi'_i$ ($i=1, 2, \dots, k$) Δ vanishes. Also when Δ vanishes, we have conversely $\pi_i = \pi'_i$ ($i=1, 2, \dots, k$) that is, the two populations are identical.

3. DIVERGENCE BETWEEN A SAMPLE AND THE POPULATION, ITS RELATION TO PEARSONIAN χ^2 FOR LARGE n

Let n be the size of a sample and (p_i) be the observed proportions taken from a population with proportions (π_i) . Then, from the definition of divergence, the divergence β^2 of the sample from the population is given by

$$\begin{aligned} \cos \beta &= \sum_{i=1}^k \sqrt{p_i \pi_i} = \sum_{i=1}^k \pi_i \left[1 + \frac{p_i - \pi_i}{\pi_i} \right]^{\frac{1}{2}} \quad \dots (3.1) \\ &= \sum_{i=1}^k \pi_i \left[1 + \frac{p_i - \pi_i}{2\pi_i} - \frac{1}{2} \frac{(p_i - \pi_i)^2}{\pi_i^2} + \epsilon_i \right] \end{aligned}$$

where $|\epsilon_i|$ is of the order of $\frac{1}{n^{3/2}}$

$$\begin{aligned} &= \sum_{i=1}^k \pi_i + \frac{1}{2} \sum_{i=1}^k (p_i - \pi_i) - \frac{1}{2} \sum_{i=1}^k \frac{(p_i - \pi_i)^2}{\pi_i} + \sum_{i=1}^k \epsilon_i \pi_i \\ &= 1 - \frac{1}{2} \frac{\chi^2}{n} + \epsilon \quad \dots (3.2) \end{aligned}$$

where $\chi^2 = \sum(n p_i - n \pi_i)^2 / n \pi_i$, the Pearsonian χ^2 and ϵ is $O(1/n^{1/2})$. Thus

$$2 \sin^2 \frac{\beta}{2} = \frac{1}{2} \frac{\chi^2}{n} \quad \dots (3.3)$$

approximately, when n is very large. Now we know that, for large values of k , $(\sqrt{2}\chi^2 - \sqrt{2k-3})$ is distributed normally with unit standard deviation. So, χ^2 is of the order of k and χ^2/n is very small.

So, finally we get

$$\beta^2 = \frac{1}{2} \frac{\chi^2}{n} \text{ i.e. } \beta = \frac{1}{\sqrt{2}} \frac{\chi}{n} \quad \dots (3.4)$$

4. ASYMPTOTIC FORM OF THE MULTINOMIAL PROBABILITY WHEN n IS LARGE

We know that the multinomial probability

$$\frac{n!}{(n p_1)! (n p_2)! \dots (n p_k)!} \pi_1^{n p_1} \pi_2^{n p_2} \dots \pi_k^{n p_k}$$

$$\sim \text{Const. } e^{-\chi^2/2} \frac{d p_1}{\sqrt{\pi_1}} \frac{d p_2}{\sqrt{\pi_2}} \dots \frac{d p_k}{\sqrt{\pi_k}}$$

$$\sim \text{Const. } e^{-2n\beta^2} d\sqrt{p_1} d\sqrt{p_2} \dots d\sqrt{p_k}$$

$$\text{for } \sqrt{p_i} = \sqrt{\pi_i + (p_i - \pi_i)} \left[\frac{d\sqrt{\pi_i}}{d\pi_i} \right] + \frac{(p_i - \pi_i)^2}{2!} \left[\frac{d^2\sqrt{\pi_i}}{d\pi_i^2} \right] + \dots$$

(by Taylor's theorem) and so $d\sqrt{p_i} = d p_i / \sqrt{\pi_i}$ approximately, for $\pi_i \neq 0$ and $(p_i - \pi_i)$ is $O(1/n^{1/2})$

$$\sim \text{Const. } e^{-2n\beta^2} d a \quad \dots (4.1)$$

where $d a$ is an elementary volume given by $d\sqrt{p_1} d\sqrt{p_2} \dots d\sqrt{p_k}$ subject to $\sum p = 1$.

The above could be seen to be true also from the following considerations. We know that the multinomial distribution may be considered to be composed of a number of Poisson distributions.

Thus

$$\frac{n!}{(n p_1)! (n p_2)! \dots (n p_k)!} \pi_1^{n p_1} \pi_2^{n p_2} \dots \pi_k^{n p_k}$$

$$= \frac{n!}{e^n \cdot n^n} \frac{e^{-n\pi_1} \pi_1^{n p_1}}{(n p_1)!} \cdot \frac{e^{-n\pi_2} \pi_2^{n p_2}}{(n p_2)!} \dots \frac{e^{-n\pi_k} \pi_k^{n p_k}}{(n p_k)!}$$

We also know (Bartlett: 1933) that in Poisson's distribution if x is the observed and a the expected value, then \sqrt{x} is distributed about \sqrt{a} normally with standard deviation $1/2$. The above expression becomes

$$\text{Const. } e^{-2n(\sqrt{p_1} - \sqrt{\pi_1})^2} d\sqrt{p_1} \cdot e^{-2n(\sqrt{p_2} - \sqrt{\pi_2})^2} d\sqrt{p_2} \dots e^{-2n(\sqrt{p_k} - \sqrt{\pi_k})^2} d\sqrt{p_k}$$

$$= \text{Const. } e^{-2n \sum_{i=1}^k (\sqrt{p_i} - \sqrt{\pi_i})^2} d\sqrt{p_1} \cdot d\sqrt{p_2} \dots d\sqrt{p_k}$$

$$\sim \text{Const. } e^{-2n \cdot 4 \sin^2 \frac{\beta}{2}} d a$$

$$\sim \text{Const. } e^{-2n\beta^2} d a \quad \dots (4.2)$$

be same as (4.1), on the supposition that β is very small.

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5. GEOMETRICAL INTERPRETATION OF THE VOLUME ELEMENT

As the point $(\sqrt{p_1} \dots \sqrt{p_k})$ lies on the unit hypersphere in the k -dimensional space, $d\sqrt{p_1} \dots d\sqrt{p_k}$ is an elementary surface volume of a sphere round the sample point, say S . With this geometrical representation, the probability of a sample is completely representable on the unit hypersphere. Assuming the law to hold for all values of β , we may find the distribution of β . For keeping β fixed if we integrate out over the available space on the sphere, we get something like a zone of annular space. The volume of this space is proportional to β^{2k-2} (for the volume of the zone is proportional to $\sin^{2k-2}\beta$ and so approximately to $\beta^{2k-2}d\beta$). So, the distribution of β is

$$\text{Const. } e^{-2n\beta^2} \beta^{2k-2} d\beta \quad \dots (5.1)$$

which is also obvious from the relation between β^2 and χ^2 .

6. DEFINITION OF DIVERGENCE BETWEEN TWO SAMPLES

Corresponding to Δ^2 in the populations, we have the statistic defined by

$$\text{cos } D' = \sqrt{p_1 p_1' + \sqrt{p_2 p_2' + \dots + \sqrt{p_k p_k'}}$$

where n, n' are the sizes and (p_i) and (p_i') are the sample proportions. This is not an unbiased estimate of Δ^2 .

7. RELATION OF D'^2 WITH PEARSONIAN χ^2 WHEN $\Delta=0$ AND n AND n' ARE LARGE AND EQUAL TO ONE ANOTHER

When the sample sizes are the same, namely n , we can express D'^2 in terms of Pearsonian χ^2 . For

$$\begin{aligned} \text{cos } D' &= \sum_{i=1}^k \sqrt{p_i p_i'} = \frac{1}{2} \sum_{i=1}^k \sqrt{(p_i - p_i')^2 + (p_i + p_i')^2} \\ &= \frac{1}{2} \sum_{i=1}^k (p_i + p_i') \left\{ 1 - \frac{(p_i - p_i')^2}{(p_i + p_i')^2} \right\}^{\frac{1}{2}} \\ &= \frac{1}{2} \sum_{i=1}^k (p_i + p_i') \left\{ 1 - \frac{1}{2} \frac{(p_i - p_i')^2}{(p_i + p_i')^2} - \frac{1}{8} \frac{(p_i - p_i')^4}{(p_i + p_i')^4} - \dots \right\} \end{aligned}$$

by the binomial theorem, because $\frac{(p_i - p_i')}{(p_i + p_i')} < 1$,

$$\begin{aligned} &= \frac{1}{2} \sum_{i=1}^k (p_i + p_i') - \frac{1}{4} \sum_{i=1}^k \frac{(p_i - p_i')^2}{(p_i + p_i')} - \frac{1}{8} \sum_{i=1}^k \epsilon_i \\ &= 1 - \frac{1}{2} \sum_{i=1}^k \frac{(p_i - p_i')^2}{(p_i + p_i')} - \epsilon \quad \dots (7.1) \end{aligned}$$

Now, Δ^2 being zero, $(p_i - p_i')$ is of the order of $1/\sqrt{n}$ and so ϵ is of the order of $1/n^2$. So, when n is very large,

$$2 \sin^2 \frac{D'}{2} = \frac{1}{2} \sum_{i=1}^k \frac{(p_i - p_i')^2}{p_i + p_i'} \quad \dots (7.2)$$

The sample values are $n p_i$ and $n p'_i$ ($i=1, 2, \dots, k$). So from the usual method of calculating χ^2 we get

$$\begin{aligned} \chi^2 &= \sum_{i=1}^k \frac{\left(n p_i - \frac{n p_i + n p'_i}{2} \right)^2}{\frac{n p_i + n p'_i}{2}} + \sum_{i=1}^k \frac{\left(n p'_i - \frac{n p_i + n p'_i}{2} \right)^2}{\frac{n p_i + n p'_i}{2}} \\ &= \sum_{i=1}^k \frac{(n p_i - n p'_i)^2}{n p_i + n p'_i} \quad \dots (7.3) \end{aligned}$$

This χ^2 , we know, is of the order of $2k$. So, χ^2/n is small and as from (7.2) $2\epsilon \sin^2 D'/2 = \chi^2/4n$ we get $D'^2 = \chi^2/2n$. From this relation it is evident that on the null hypothesis, the distribution of D'^2 follows the distribution of χ^2 for very large samples.

8. DISTRIBUTION OF D'^2 WHEN $\Delta^2 \neq 0$

Let there be two samples with sizes n and n' and proportions (p_i) and (p'_i) taken respectively from the two populations characterised by (π_i) and (π'_i). Let β^2 and β'^2 be the divergences (i.e. square of the angular distances) of the two samples from their respective populations. Then the joint distribution of the two samples is

$$\text{Const. } e^{-2\beta n^2} d\alpha \cdot e^{-2\beta' n'^2} d\alpha' \quad \dots (8.1)$$

where $d\alpha$ and $d\alpha'$ are elementary surface volumes, on the unit hypersphere, about the sample points. Let P and P' be the population points (i.e. where the population lines meet the unit sphere), S and S' the sample points (with the same convention as the population points). Then Δ is represented by PP' , D' by SS' , and β and β' by SP and $S'P'$. From the distributions of β and β' we know, when n and n' are very large, that these tend to become very small. That is, the probability of a large β is so small that, for all practical purposes, β may be regarded to lie within a narrow zone about the arc PP' . Similar is the case for β' . Outside this zone the probability of getting samples is negligible. This narrow zone may be conceived to be a portion of a cylinder (that is, the projection of this zone on the enveloping cylinder osculating the unit sphere along the curve PP' may be regarded to be coincident with the zone itself). We may develop out this cylindrical zone into a flat of $(k-1)$ dimensions where the geodesics β , β' , Δ and D' all become right lines and $d\alpha$ and $d\alpha'$ become elementary areas (i.e. volumes) in this flat of $(k-1)$ dimensions.

With the help of the above geometrical device we may proceed to find the distribution of D' which now represents a rectilinear length. In the $(k-1)$ flat space let the coordinates of the points P , P' , S and S' be (α_i) , (α'_i) , (x_i) and (x'_i) , respectively ($i=1, 2, \dots, k-1$). The joint probability of the two samples is

$$\begin{aligned} &\text{const. } e^{-2n\beta^2 - 2n'\beta'^2} d\alpha d\alpha' \\ &= \text{const. } e^{-2n \sum_{i=1}^{k-1} (x_i - \alpha_i)^2 - 2n' \sum_{i=1}^{k-1} (x'_i - \alpha'_i)^2} \prod_{i=1}^{k-1} dx_i \prod_{i=1}^{k-1} dx'_i \end{aligned}$$

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From this it is easy to deduce the distribution of $(x_1 - x'_1 - a_1 + a'_1)$ in the form

$$\text{Const. } e^{-\frac{2n n'}{n+n'} \left\{ \sum_{i=1}^{k-1} (x_i - x'_i - a_i + a'_i)^2 \right\}} \prod_{i=1}^{k-1} d(x_i - x'_i)$$

Or, which is the same as

$$\text{Const. } e^{-\frac{2n n'}{n+n'} \left\{ \sum_{i=1}^{k-1} (x_i - x'_i)^2 - 2 \sum_{i=1}^{k-1} (x_i - x'_i)(a_i - a'_i) + \sum_{i=1}^{k-1} (a_i - a'_i)^2 \right\}} \prod_{i=1}^{k-1} d(x_i - x'_i)$$

From this by adopting the method employed by Mr. R. C. Bose for getting the distribution of the classical D^2 (Bose 1936) we get from the above

$$\text{Const. } e^{-\frac{2n n'}{n+n'} \left\{ D^2 - 2D'\Delta \cos \theta + \Delta^2 \right\}} (D' \sin \theta)^{k-1} D' d\theta$$

Where $D^2 = \sum_{i=1}^{k-1} (x_i - x'_i)^2$, $\Delta^2 = \sum_{i=1}^{k-1} (a_i - a'_i)^2$ and $D'\Delta \cos \theta = \sum_{i=1}^{k-1} (x_i - x'_i)(a_i - a'_i)$

Integrating over θ from 0 to 2π we get the distribution of D as

$$\text{Const. } D^{k-1} e^{-\frac{2n n'}{n+n'} (D^2 + \Delta^2)} \int_0^{2\pi} \frac{4n n'}{n+n'} D'\Delta d\theta$$

where \int is a Bessel function of purely imaginary argum. nt.

9. APPROXIMATION OF Δ AND D' WHEN Δ IS VERY SMALL

We know

$$\cos \Delta = \sum \sqrt{n_i n'_i} = \frac{1}{2} \sum (n_i + n'_i) \left[1 - \left(\frac{n_i - n'_i}{n_i + n'_i} \right)^2 \right]^{\frac{1}{2}}$$

Replacing $\cos \Delta$ by $1 - \Delta^2/2$ to a first approximation we have $\Delta^2 = \sum (n_i - n'_i)^2 / (n_i + n'_i)$. Similarly $D'^2 = \sum (p_i - p'_i)^2 / (p_i + p'_i)$. These were suggested by Prof. Mahalanobis as a measure of divergence between two multinomial populations and samples therefrom.

10. SPECIAL CASE OF BINOMIAL POPULATIONS

Let there be two binomial populations defined by (n_1, n_2) and (n'_1, n'_2) from which two samples of sizes n and n' with proportions (p_1, p_2) and (p'_1, p'_2) are taken respectively. Then we know that $n(p_1 - p_2) / \sqrt{n p_1 (1 - p_1)}$ is distributed normally about zero with unit standard deviation subject to the condition that n is sufficiently large and both n_1 and n_2 not far different from $n/2$. Now, under the same assumptions, we have $(\sin^{-1} \sqrt{p_1} - \sin^{-1} \sqrt{p_2}) = (p_1 - p_2) / 2\sqrt{p_1(1-p_1)}$ approximately (by expanding in Taylor's series). Therefore $2\sqrt{n}(\sin^{-1} \sqrt{p_1} - \sin^{-1} \sqrt{p_2})$ is distributed normally about zero with unit standard deviation. Similarly, under the same set of conditions, $2\sqrt{n'}(\sin^{-1} \sqrt{p'_1} - \sin^{-1} \sqrt{p'_2})$ is distributed normally with unit standard deviation. So, from these two it easily follows that

$$2\sqrt{\frac{n n'}{n+n'}} \left\{ (\sin^{-1} \sqrt{p_1} - \sin^{-1} \sqrt{p'_1}) - (\sin^{-1} \sqrt{p_2} - \sin^{-1} \sqrt{p'_2}) \right\}$$

is distributed with unit standard deviation. Here the population divergence is clearly $\sin^{-1}\sqrt{p_1} - \sin^{-1}\sqrt{p_1'}$ and the sample estimate is $\sin^{-1}\sqrt{p_1} - \sin^{-1}\sqrt{p_1'}$. Clearly this is a particular case of the general problem which degenerates into it when $k=2$.

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