NOTES

This section is devoted to brief research and expository articles and other short items.

THE DISTRIBUTION OF STUDENT'S $\,t\,$ WHEN THE POPULATION MEANS ARE UNEQUAL

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Let x_1, \dots, x_N be independent normal variates with the same variance σ^2 and with means μ_1, \dots, μ_N respectively. Set n = N - 1 and let

(1)
$$\bar{x} = \sum_{i=1}^{N} x_i/N, \quad s^2 = \sum_{i=1}^{N} (x_i - \bar{x})^2/n, \quad t = N^{\frac{1}{2}} \bar{x}/s.$$

If all the μ_i are 0 then t has Student's distribution with n degrees of freedom; its frequency function will be denoted here by

(2)
$$f_{n,0}(t) = n^{-\frac{1}{2}} \left[B\left(\frac{1}{2}, \frac{n}{2}\right) \right]^{-1} \cdot (1 + t^2/n)^{-\frac{1}{2}(n+1)}.$$

When dealing with situations involving mixtures of populations or in which the mean exhibits a secular trend, it is important to know the distribution of t when the μ_i are arbitrary; in the general case let

(3)
$$\bar{\mu} = \sum_{1}^{N} \mu_{i}/N, \qquad \beta^{2} = \sum_{1}^{N} (\mu_{i} - \bar{\mu})^{2}/N,$$
$$\alpha = N\bar{\mu}^{2}/2\sigma^{2}, \qquad \lambda = N\beta^{2}/2\sigma^{2}.$$

The distribution of t will be shown to depend on the three parameters n, α , λ . If $\lambda = \beta^2 = 0$, so that all the μ_i are equal, then the distribution of t determines the power function of the ordinary t test. We shall here consider the case in which $\alpha = \bar{\mu} = 0$, although the μ_i are different. Denoting the frequency function of t in this case by $f_{n,\lambda}(t)$ we shall show that

(4)
$$f_{n,\lambda}(t) = f_{n,0}(t) \cdot \exp\left\{\frac{-\lambda t^2/n}{1+t^2/n}\right\} \cdot F(-\frac{1}{2}, n/2, -\lambda(1+t^2/n)^{-1}),$$

where F denotes the confluent hypergeometric series, and where, since $\bar{\mu} = 0$,

$$\lambda = \sum_{i=1}^{N} \mu_i^2 / 2\sigma^2.$$

In fact, the general distribution of t, of which (4) represents the case $\alpha = 0$,

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may be derived as follows. Using the standard orthogonal transformation [1, p. 387] let

(6)
$$z_i = \sum_{i=1}^{N} c_{ij} x_j, \qquad x_i = \sum_{i=1}^{N} c_{ji} z_j \qquad (i = 1, \dots, N),$$

where

(7)
$$c_{1j} = N^{-\frac{1}{2}}$$
 $(j = 1, \dots, N);$

then

(8)
$$t = n^{\frac{1}{2}} z_1 / \left(\sum_{i=1}^{N} z_i^2\right)^{\frac{1}{2}}.$$

The joint frequency function of the z_i is easily seen to be

(9)
$$(2\pi)^{-N/2} \cdot \sigma^{-N} \cdot \exp\left\{-\sum_{i=1}^{N} (z_{i} - a_{i})^{2}/2\sigma^{2}\right\},$$

where

(10)
$$a_1 = N^{\frac{1}{2}}\bar{\mu}, \qquad \sum_{i=1}^{N} a_i^2 = N\beta^2.$$

Thus t is the ratio of a non-central normal variate to the square root of an independent non-central chi-square variate. It is known [2, p. 138] that the frequency function of $q^2 = \sum_{n=0}^{N} z_i^2/\sigma^2$ is

(11)
$$\frac{1}{2}e^{-\lambda} \cdot (\frac{1}{2}q^2)^{\frac{1}{2}n-1} \cdot e^{-q^2/2} \cdot \sum_{j=0}^{\infty} \frac{(\frac{1}{2}\lambda q^2)^j}{j! \Gamma(\frac{1}{2}n+j)},$$

where

(12)
$$\lambda = \sum_{i=1}^{N} a_{i}^{2}/2\sigma^{2} = N\beta^{2}/2\sigma^{2}.$$

The frequency function of $v = z_1/\sigma$ is

$$g(v) = \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{(\sigma v - a_1)^2}{2\sigma^2}\right\} = \frac{1}{\sqrt{2\pi}} e^{-\alpha} \cdot e^{-(v^2/2)} \cdot \sum_{k=0}^{\infty} \frac{(2\alpha)^{k/2}}{k!} x^k,$$

that of q is, by (11),

$$h(q) = 2^{1-(n/2)}e^{-\lambda}e^{-(q^2/2)} \sum_{j=0}^{\infty} \frac{\lambda^j q^{2j+n-1}}{2^{j}j!\Gamma((n/2)+j)}, \qquad (q > 0),$$

hence that of $u = v/q = n^{-\frac{1}{2}}t$ is

$$\int_0^\infty h(q)g(uq)q\,dq,$$

which, after integration, reduces to

(13)
$$\pi^{-\frac{1}{2}}e^{-(\lambda+\alpha)}\sum_{j=0}^{\infty}\sum_{k=0}^{\infty}\frac{\lambda^{j}(2\alpha^{\frac{1}{2}}u)^{k}}{j!\,k!}\frac{\Gamma(N/2+j+k/2)}{\Gamma(n/2+j)}(1+u^{2})^{-(N+2j+k)/2}.$$

In particular, if $\alpha = \bar{\mu} = 0$ then (13) reduces by means of the relation $F(\alpha, \gamma, x) = e^x F(\gamma - \alpha, \gamma, -x)$ to

$$(14) \quad \left[B\left(\frac{1}{2},\frac{n}{2}\right)\right]^{-1} \cdot e^{-\lambda u^2/(1+u^2)} \cdot (1+u^2)^{-\frac{1}{2}N} \cdot F\left(-\frac{1}{2},\frac{n}{2},-\lambda(1+u^2)^{-1}\right),$$

from which it follows that the frequency function of t is given by (4).

Again, let x_1 , \cdots , $x_{N_1+N_2}$ be independent normal variates with the same variance σ^2 and with means μ_1 , \cdots , $\mu_{N_1+N_2}$ respectively. Set $n_1 = N_1 - 1$, $n_2 = N_2 - 1$, $n_2 = n_1 + n_2$, and let

$$\bar{x}_{1} = \sum_{1}^{N_{1}} x_{i}/N_{1}, \qquad \bar{x}_{2} = \sum_{N_{1}+1}^{N_{1}+N_{2}} x_{i}/N_{2}$$

$$(15) \qquad s_{1}^{2} = \sum_{1}^{N_{1}} (x_{i} - \bar{x}_{1})^{2}/n_{1}, \qquad s_{2}^{2} = \sum_{N_{1}+1}^{N_{1}+N_{2}} (x_{i} - \bar{x}_{2})^{2}/n_{2}$$

$$s^{2} = (n_{1}s_{1}^{2} + n_{2}s_{2}^{2})/(n_{1} + n_{2}), \quad t = [N_{1}N_{2}/(N_{1} + N_{2})]^{\frac{1}{2}}(\bar{x}_{1} - \bar{x}_{2})/s.$$

If all the μ_i are equal then t again has Student's distribution with n degrees of freedom. In the general case let

(16)
$$\bar{\mu}_{1} = \sum_{1}^{N_{1}} \mu_{i}/N_{1}, \quad \bar{\mu}_{2} = \sum_{N_{1}+1}^{N_{1}+N_{2}} \mu_{i}/N_{2},$$

$$\beta_{1}^{2} = \sum_{1}^{N_{1}} (\mu_{i} - \bar{\mu}_{1})^{2}/N_{1}, \quad \beta_{2}^{2} = \sum_{N_{1}+1}^{N_{1}+N_{2}} (\mu_{i} - \bar{\mu}_{2})^{2}/N_{2}.$$

Then we may show as before [1, p. 388] that in this case $u = n^{-\frac{1}{2}}t$ has the frequency function (13), where now

(17)
$$N = N_1 + N_2 - 1, \quad \lambda = (N_1 \beta_1^2 + N_2 \beta_2^2)/2\sigma^2,$$
$$\alpha = [N_1 N_2/(N_1 + N_2)](\bar{\mu}_1 - \bar{\mu}_2)^2/\sigma^2.$$

In particular, when $\alpha = \bar{\mu}_1 - \bar{\mu}_2 = 0$, so that $\bar{\mu}_1 = \bar{\mu}_2 = \bar{\mu}$, say, the frequency function $f_{n,\lambda}(t)$ of t is again given by (4), where now

(18)
$$\lambda = \sum_{1}^{N_1+N_2} (\mu_i - \bar{\mu})^2 / 2\sigma^2.$$

Extensions in this direction to the general linear hypothesis in the analysis of variance will not be treated here

If we set

$$(19) w = (1 + t^2/n)^{-1}$$

where t has the frequency function (4), then w will have the frequency function

$$(20) \ g_{n,\lambda}(w) = \left[B\left(\frac{1}{2}, \frac{n}{2}\right) \right]^{-1} \cdot e^{-\lambda(1-w)} \cdot w^{\frac{1}{2}n-1} \cdot (1-w)^{-\frac{1}{2}} \cdot F\left(-\frac{1}{2}, \frac{n}{2}, -\lambda w\right),$$

for $0 < w \le 1$. Thus for every t,

(21)
$$1 - \int_{-t}^{t} f_{n,\lambda}(x) \ dx = \int_{0}^{(1+t^{2}/n)^{-1}} g_{n,\lambda}(w) \ dw_{\bullet}$$

It would be interesting to have numerical values of the integral on the left side of (21) for that value of t for which

(22)
$$1 - \int_{-t}^{t} f_{n,0}(x) \ dx = 0.01 \text{ or } 0.05 \text{ (say)},$$

but existing tables (e.g. those in [2] and [3]) of the integral of (20) were compiled for a different purpose and do not supply this information. The following remarks throw some light on this subject.

Let us set

$$R(t) = f_{n,\lambda}(t)/f_{n,0}(t) = \exp\left\{\frac{-\lambda t^2/n}{1+t^2/n}\right\} \cdot F\left(-\frac{1}{2}, \frac{n}{2}, -\lambda(1+t^2/n)^{-1}\right)$$

$$= \left\{1 - \lambda(t^2/n)/(1+t^2/n) + o(\lambda)\right\}$$

$$\cdot \left\{1 + \lambda/(n+t^2) + o(\lambda)\right\}$$

$$= 1 + \lambda(n+t^2)^{-1}(1-t^2) + o(\lambda).$$

Then as $\lambda \to 0$ we have ultimately

(24)
$$R(t) > 1 \text{ if } |t| < 1,$$

$$R(t) < 1 \text{ if } |t| > 1.$$

Hence for any t > 1 and for sufficiently small λ ,

(25)
$$1 - \int_{-t}^{t} f_{n,\lambda}(x) \ dx < 1 - \int_{-t}^{t} f_{n,0}(x) \ dx.$$

The exact range of values of t for which R(t) < 1 depends of course on n and λ . However we shall show that always

(26)
$$R(t) < 1 \text{ if } |t| > 1,$$

so that (25) holds for all n and $\lambda > 0$, provided t > 1. The proof is as follows. In terms of w we have

(27)
$$R(t) = e^{-\lambda(1-w)} \cdot F(-\frac{1}{2}, n/2, -\lambda w) = e^{-\lambda} F((n+1)/2, n/2, \lambda w).$$

Now

(28)
$$F((n+1)/2, n/2, \lambda w) = 1 + \sum_{k=1}^{\infty} \frac{(n+1)(n+3)\cdots(n+2k-1)}{n(n+2)\cdots(n+2k-2)} (\lambda w)^k / k!,$$

and by induction on k we may show that for all $k = 1, 2, \dots$,

(29)
$$\frac{(n+1)(n+3)\cdots(n+2k-1)}{n(n+2)\cdots(n+2k-2)} \le 1 + k/n,$$

where the equality holds only for k = 1. Hence

(30)
$$F((n+1)/2, n/2, \lambda w) < 1 + \sum_{k=1}^{\infty} (1 + k/n) \cdot (\lambda w)^k / k! = e^{\lambda w} (1 + \lambda w/n),$$

(31)
$$R(t) < e^{-\lambda(1-w)} \cdot (1 + \lambda w/n) < e^{-\lambda(1-w)} \cdot e^{\lambda w/n} = e^{-\lambda[1-w(1+1/n)]}$$

Hence R(t) < 1 if w < n/(n+1), which is equivalent to (26).

REFERENCES

- [1] H. Cramér, Mathematical Methods of Statistics, Princeton University Press, Princeton, 1946.
- [2] P. C. Tang, "The power function of the analysis of variance tests with tables and illustrations of their use," Stat. Res. Memoirs, Vol. 2 (1938), pp. 127-149.
- [3] Emma Lehmer, "Inverse tables of probabilities of errors of the second kind," Annals of Math. Stat., Vol. 15 (1944), pp. 388-398.

A DISTRIBUTION-FREE CONFIDENCE INTERVAL FOR THE MEAN

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1. Summary. Consider a random sample of N observations x_1, x_2, \dots, x_N , from a universe of mean μ and variance σ^2 . Let m and s^2 be the sample mean and variance respectively:

(1)
$$m = \frac{1}{N} \sum_{i=1}^{N} x_i, \qquad s^2 = \frac{1}{N} \sum_{i=1}^{N} (x_i - m)^2.$$

It is shown that the following conservative confidence interval holds for '\mu:

(2) Prob
$$\{(m-\mu)^2 \le s^2/(N-1) + \lambda \sigma^2 \sqrt{2/N(N-1)}\} > 1 - \lambda^{-2}$$
,

where λ is any positive constant. Inequality (2) also holds if, in the braces, λ is replaced by $\sqrt{\lambda^2 - 1}$, with $\lambda \ge 1$.

Inequality (2) is much more efficient on the average than Tchebychef's inequality for the mean, namely,

(3) Prob
$$\{(m-\mu)^2 \le \lambda^2 \sigma^2/N\} > 1 - \lambda^{-2},$$

yet (2) and (3) are both distribution-free, requiring only knowledge about σ^2 . At the $1 - \lambda^{-2} = .99$ level of confidence, the expected value of the right member in the braces of (2) is only about 1/6 the corresponding member of (3); at the .999 level of confidence the ratio is about 1/20.