

**On the characterisation of the normal population by
the independence of the sample mean
and the sample variance.**

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1. Let X_1, X_2, \dots, X_n ($n \geq 2$) be the sample variables from a certain population, that is, let X_i ($i=1, 2, \dots, n$) be independent random variables having same distribution $F(x)$. In the mathematical statistics, the following fact is well known and is of fundamental importance in the theory of exact sampling.

If $F(x)$ is the normal distribution function, then the two statistics

$$(1.1) \quad \bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

and

$$(1.2) \quad S = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$$

are statistically independent.

R. C. Geary⁽¹⁾ has proved the converse of this theorem and given the characterisation of the normal population by using the formulae⁽²⁾ due to Fisher for relations between semi-invariants of various algebraic forms of sample variables. The object of the present paper is to give another proof, under the more general conditions assuming nothing about the moments of X_i , while Geary has supposed the existence of moments of every order.

2. We restate the theorem.

Theorem. Let X_1, X_2, \dots, X_n ($n \geq 2$) be the independent random variables whose distributions are equal to the same $F(x)$. If two random variables $Y = \sum_{i=1}^n X_i$, $Z = \sum_{i=1}^n (X_i - \bar{X})^2$ ($\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$) are independently distributed,

(1) R. C. Geary, The distribution of "Student's" ratio for non-normal samples. Journ. Royal Statist. Soc., Supplement 3 (1936).

(2) R. A. Fisher, Moments and product moments of sampling distributions. Proc. London Math. Soc. 30 (1929).

then $F(x)$ must be the normal distribution function, excluding the unit distribution.

We consider the characteristic function⁽³⁾ of the simultaneous variable (X_i, X_i^2)

$$(2.1) \quad f(t, s) = \int_{-\infty}^{\infty} e^{itx + isx^2} dF(x),$$

where t is a real number but we consider s as a complex number, $s = \sigma + i\tau$, $\tau > 0$. $f(t, s)$ is obviously an analytic function of s regular in the upper half-plane $\tau > 0$. Since $X_i (i=1, 2, \dots, n)$ are independent variables, the characteristic function of variable (Y, Σ) , Σ being $\sum_{i=1}^n X_i^2$, is $\{f(t, s)\}^n$ which noticing that $\Sigma \geq 0$, can also be written as

$$(2.2) \quad \int_{-\infty}^{\infty} \int_0^{\infty} e^{it\eta + is\theta} dF(\eta, \theta),$$

where $F(\eta, \theta)$ is the distribution function of (Y, Σ) .

Since $Z + \frac{1}{n} Y^2 = \Sigma$, denoting the distribution of (Y, Z) as $G(\eta, \zeta)$, we have further

$$(2.3) \quad \{f(t, s)\}^n = \int_{-\infty}^{\infty} \int_0^{\infty} e^{it\eta + is(\frac{\eta^2}{n} + \zeta)} dG(\eta, \zeta).$$

The statistical independence of Y and Z shows that

$$(2.4) \quad dG(\eta, \zeta) = dG_1(\eta) dG_2(\zeta),$$

$G_1(\eta)$ and $G_2(\zeta)$ being the distribution function of Y and Z respectively. Hence we can write (2.3) as

$$(2.5) \quad \{f(t, s)\}^n = \int_{-\infty}^{\infty} e^{it\eta + i\frac{s}{n}\eta^2} dG_1(\eta) \cdot \int_0^{\infty} e^{is\zeta} dG_2(\zeta).$$

Now we observe that, putting $\phi\left(t, \frac{s}{n}\right) = \int_{-\infty}^{\infty} e^{it\eta + i\frac{s}{n}\eta^2} dG_1(\eta)$ and $u(s) = u_n(s) = \int_0^{\infty} e^{is\zeta} dG_2(\zeta)$,

(3) In the ordinary sense of the characteristic function, t, s are real, but in this paper, we use the same terminology in the case where s is complex.

$$(2.6) \quad -i \frac{\partial^2}{\partial t^2} f(t, s) = \frac{\partial}{\partial s} f(t, s), \quad \tau > 0,$$

$$(2.7) \quad -\frac{i}{n} \frac{\partial^2}{\partial t^2} \psi\left(t, \frac{s}{n}\right) = \frac{\partial}{\partial s} \psi\left(t, \frac{s}{n}\right), \quad \tau > 0,$$

and

$$(2.8) \quad -ia'(s) \geq 0 \quad \text{for } \sigma=0, \tau > 0.$$

We differentiate both sides of (2.5) with respect to s , we have

$$n \{f(t, s)\}^{n-1} \frac{\partial}{\partial s} f(t, s) = a(s) \frac{\partial}{\partial s} \psi\left(t, \frac{s}{n}\right) + \psi\left(t, \frac{s}{n}\right) a'(s). \quad (4)$$

which becomes, by (2.6) and (2.7),

$$(2.9) \quad n \{f(t, s)\}^{n-1} \frac{\partial^2}{\partial t^2} f(t, s) = \frac{1}{n} a(s) \frac{\partial^2}{\partial t^2} \psi\left(t, \frac{s}{n}\right) + i \psi\left(t, \frac{s}{n}\right) a'(s).$$

In differentiating two times (2.5) with respect to t we get

$$(2.10) \quad n \{f(t, s)\}^{n-1} \frac{\partial^2}{\partial t^2} f(t, s) + n(n-1) \{f(t, s)\}^{n-1} \left\{ \frac{\partial}{\partial t} f(t, s) \right\}^2 \\ = a(s) \frac{\partial^2}{\partial t^2} \psi\left(t, \frac{s}{n}\right).$$

The elimination of $a(s) \frac{\partial^2}{\partial t^2} \psi\left(t, \frac{s}{n}\right)$ from (2.9) and (2.10) gives

$$\{f(t, s)\}^{n-1} \frac{\partial^2}{\partial t^2} f(t, s) - \{f(t, s)\}^{n-1} \left\{ \frac{\partial}{\partial t} f(t, s) \right\}^2 = i \frac{a'}{n-1} \psi\left(t, \frac{s}{n}\right).$$

By (2.5), this becomes further

$$(2.11) \quad \{f(t, s)\}^{n-1} \frac{\partial^2}{\partial t^2} f(t, s) - \{f(t, s)\}^{n-2} \left\{ \frac{\partial}{\partial t} f(t, s) \right\}^2 \\ = i \{f(t, s)\}^n \frac{1}{n-1} \frac{a'(s)}{a(s)}.$$

From this equation we can easily prove that in the t -interval, for fixed s , such that $f(t, s) \neq 0$,

(4) The dash in $a'(s)$ means the differentiation with respect to s .

$$(2.12) \quad f(t, s) = \exp \left[\frac{i}{n-1} \frac{a'(s)}{a(s)} \left\{ \frac{t^2}{2} + C(s)t + D(s) \right\} \right].$$

But since $f(t, s)$ is a continuous function of t and the right side of (2.12) has no zeros as a function of t , we see that (2.12) holds for all values of t .

Now we take $\sigma=0$, and thus $s=i\tau$. Then it holds that

$$(2.13) \quad \lim_{\tau \rightarrow +0} f(t, i\tau) = \int_{-\infty}^{\infty} e^{itx} dF(x),$$

for every t , since

$$\begin{aligned} & \left| \int_{-\infty}^{\infty} e^{itx} dF(x) - \int_{-\infty}^{\infty} e^{itx - \tau x^2} dF(x) \right| \leq \left| \int_{-A}^A e^{it\tau} (1 - e^{-\tau x^2}) dF(x) \right| \\ & + \left| \int_{|x| > A} e^{itx} dF(x) \right| + \left| \int_{|x| > A} e^{itx - \tau x^2} dF(x) \right| \leq \tau A^2 \int_{-A}^A dF(x) + 2 \int_{|x| > A} dF(x) < \varepsilon, \end{aligned}$$

if we take A such that $2 \int_{|x| > A} dF(x) < \frac{\varepsilon}{2}$ and then take τ so small that

$$\tau A^2 \int_{-A}^A dF(x) < \frac{\varepsilon}{2}.$$

Now if we take $t=0$ in (1.12) and let τ tend to zero, then by (2.13) $f(0, i\tau) \rightarrow 1$, and hence

$$\lim_{\tau \rightarrow 0} \frac{a'(i\tau)}{a(i\tau)} D(i\tau) = 0.$$

Next noticing the existence of $\lim_{\tau \rightarrow 0} f(t, i\tau) f(-t, i\tau)$, ($t \neq 0$), we can show the existence of $\lim_{\tau \rightarrow 0} a'(i\tau)$. And hence we also get the existence of $\lim_{\tau \rightarrow 0} C(i\tau)$.

Let

$$\lim_{\tau \rightarrow 0} \frac{i}{n-1} \frac{a'(i\tau)}{a(i\tau)} = -a_n, \quad \lim_{\tau \rightarrow 0} \frac{i}{n-1} \frac{a'(i\tau)}{a(i\tau)} C(i\tau) = \beta_n.$$

If $a_n \neq 0$, then letting $s=i\tau \rightarrow 0$ in (2.12), we have

$$(2.14) \quad f(t) = f(t, 0) = e^{-\frac{a_n}{2} t^2 + \beta_n t}.$$

But since the left side is independent of n , a_n and β_n are constants in-

dependent of $n^{(5)}$ and thus we can put $a_n = a$, $\beta_n = i\beta$, where β is real, for $f(t) = \overline{f(-t)}$.

If $a \neq 0$; then by (2.8), $a > 0$ and (2.14) can be written as

$$f(t) = e^{-\frac{\sigma}{2}t^2 + i\beta t}$$

which shows that $F(x)$ is a normal distribution function.

If $a = 0$, then $f(t) = e^{i\beta t}$. This shows that $F(x)$ is an unit distribution function having only one point spectrum at $x = \beta$.

(5) This can also be proved explicitly. From the existence of $\lim a_n$ we can show that the variance of Z is finite, and $\frac{1}{i} \lim_{\tau \rightarrow 0} a'(i\tau) = E(Z)$, the mean value of Z , which is $(n-1)\sigma^2$, σ^2 being the variance of X_i . This is a well known fact in the sampling theory. Hence a_n is independent of n . For β_n , we can also prove its independence of n directly.