SOME HISTORICAL REFLECTIONS TRACED THROUGH THE DEVELOPMENT OF THE USE OF FREQUENCY CURVES

by

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DEPARTMENT OF STATISTICS
Southern Methodist University

## FOREWARD

Professor Egon S. Pearson of University College London visited the Department of Statistics of Southern Methodist University on March 28, 1969. He consulted with the faculty and gave a lecture to graduate students, faculty, and invited guests; this report summarizes that lecture. The same lecture was also given at several other universities in the United States.

We record here our great pleasure in having had the opportunity to meet with Professor Pearson. His spirit, humor, and gentleness are an inspiration to all who come in contact with him. The work reported on has many applications which are not yet exploited. It is our desire to make these ideas available to other potential users.

## SOME HISTORICAL REFLECTIONS, TRACED THROUGH THE DEVELOPMENT OF THE USE OF FREQUENCY CURVES\*

by

## E. S. Pearson University College London

I have chosen this subject for the theme of my talk because the changing attitude of statisticians towards frequency curves can be used - how shall I describe it? - as one of the strands which link together much of the historical development of mathematical statistics during the 50 years 1890-1940. May-be also you will like to hear something on these changes first-hand; for I came early enough into the statistical field to share in some of this development and to recapture a little of the earlier excitement of the 1890's, when the foundations of modern statistical theory were really being layed.

Let me make two points before I start: one is that when you are hearing an account of the development of ideas and methods you must not, with hind-sight, think: "Oh that should have been obvious!" One of the interesting things that one finds in studying the history of science is with what difficulty, often by trial and error, the human mind arrives at something really new.

My second point is a minor one; I shall have to refer a certain amount to my father, and as it becomes a little embarrassing to be saying frequently 'Karl Pearson' or 'my father', I shall sometimes call him more impersonally by his initials, "K.P.", a style commonly used by his contemporaries.

<sup>\*</sup>This 'lecture' is based on talks given at several statistical centres in the United States, during March and April 1969.

I suppose that my serious reading of K.P.'s early papers, mostly published in the <a href="Philosophical Transactions">Philosophical Transactions</a> of the Royal Society, began some

50 years ago, just after the end of World War I. Undoubtedly I was immensely stimulated by them, as others of an earlier generation have told me that they were. R. A. Fisher, I think, was reading these papers in the years

1912-15; he, also, was influenced and, possibly, drawn into statistics from what he found there and began to believe he could improve upon.

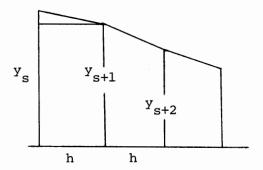
Let me start by running quickly through the early history of the statistician's fundamental distribution, the Normal curve. We know that De Moivre first used it in 1733 to provide a mathematical approximation to the binomial; he did not use it specifically as a probability distribution function of a continuous variable, but the link with the binomial is of interest because of later developments. The curve was also used by Laplace as an approximating function, but in his hands and particularly those of Gauss it formed the centre-point of the Theory of Errors. Much of this work was theoretical and it was possibly Bessel in about 1820 who was the first person actually to compare the curve with a real distribution of errors.

During the 1820's the Belgian, Adolphe Quetelet (1796-1874), who wished to establish an observatory in Brussels, put himself in touch with most of the great European mathematicians and astronomers. Rather later in life his interests turned towards demographic and vital statistics and he was able to show that the error curve also fitted certain physical measurements made on man. He was primarily interested, however, in mean values and propounded the idea of 1'homme moyen, the average man who possessed all the mean characters of his race; variations about the average he ascribed to 'accidental causes' which I do not think he attempted to explore.

the symmetrical binomial, he took a binomial with  $p=1-q\neq q$ , and derived a curve y=f(x) by equating the (slope)/(ordinate) ratio,  $\frac{1}{y}\,\frac{dy}{dx}$ , to a similar ratio found from the binomial polygon, viz. from

$$\frac{y_{s+1} - y_{s}}{h} \div \frac{1}{2} (y_{s} + y_{s+1}) .$$

Fig. 2



In this way (in the autumn of 1893) he obtained as a limit the differential equation

$$\frac{1}{y}\frac{dy}{dx} = \frac{-(x - a)}{b_0 + b_1 x} , \qquad (1)$$

with its solution

$$y = y_0 e^{-x/c} \left(1 + \frac{x}{d}\right)^m . \tag{2}$$

Next, he went further than this and applied the same treatment to the hypergeometric distribution, and reached the more general equation

$$\frac{1}{y}\frac{dy}{dx} = \frac{-(x - a)}{b_0 + b_1 x + b_2 x^2} . \tag{3}$$

Although he made some attempt, following tradition, to relate the underlying hypergeometric to a system of correlated, elementary components,

I think it may be said that to all intents and purposes he now dropped the basis of the derivation of equation (3), and started afresh from this point.

The solution of (3) (Pearson, 1895, 1901) assumed different forms according to the values of the four parameters, which could be shown to be functions of the first four moments of the distribution, y = f(x). In particular, if the variable x was standardised in the form

$$X = (x - mean)/(standard deviation)$$

the parameters in the solutions of (3) depended only on the standardised 3rd and 4th moments; these he took in the form

$$\beta_1 = \left(\frac{\mu_3}{\sigma^3}\right)^2 \quad , \quad \beta_2 = \frac{\mu_4}{\sigma^4} \quad . \tag{4}$$

Why did he use  $\beta_1$  , not  $\sqrt{\beta_1} = \mu_3/\sigma^3$  ? For practical convenience in the solution and representation.

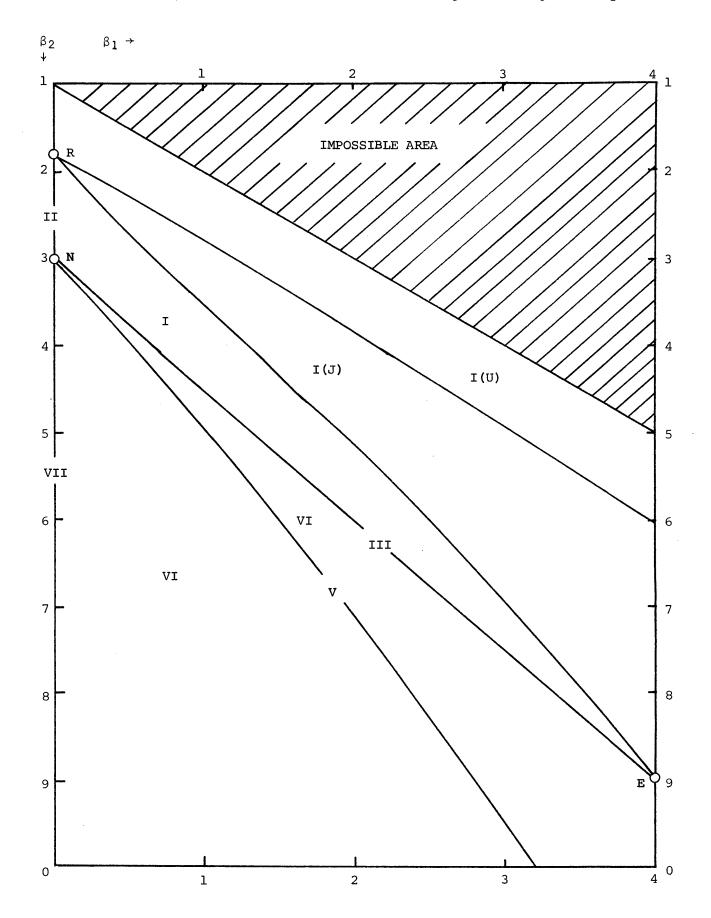
As a result, if rectangular coordinate axes  $0\beta_1$ ,  $0\beta_2$  are taken, the field is divided by certain lines into a series of regions (see Fig. 3), the bounding lines and intermediate areas being each associated with a particular form of solution to (3). The seven main types were given numbers I-VII, roughly in the order he discovered them. We might now prefer a clockwise ordering of numbers, but this would now lead to immense confusion. The result was as surprising as it was elegant. It is immaterial

- (a) whether we now choose to take the coordinates as  $\gamma_1 = \sqrt{\beta_1} \text{ and } \gamma_2 = \beta_2 3 \quad ,$
- (b) to use a new terminology, e.g., to speak of gamma, beta, inverted beta, t-distributions and so on.

The effectiveness of the system lies in this very comprehensive classification of forms of distribution, in terms of two shape parameters.

FIGURE 3. PEARSON-TYPE CURVE REGIONS IN  $\beta_1$  ,  $\beta_2$  PLANE.

Special Distributions: N=Normal; R=Uniform or Rectangular; E=Negative Exponential.



Of course there can be and are other systems. There are the Gram-Charlier or Edgeworth expansions, but if one goes only as far as the 4th moment, the Pearson system has a great advantage. It will only provide a single mode (except in the case of U-shaped distributions) and it will give no negative ordinates.

During the 1890's Pearson fitted his curves to a great variety of data, drawn from biological, medical, economic, meteorological and other sources. On the whole he obtained extremely good fits, although it was not until he developed the  $\chi^2$  test in 1900 that goodness of fit could be adequately examined. It was, perhaps, inevitable that during the next 30 to 40 years his students, in their numerical classes, had to go through the hard labour of curve fitting. I am sure there was some advantage in this hard labour, as a training in accurate computing (which meant using a hand calculating machine) and there was certainly an intellectual satisfaction in finally plotting the ordinates of the curve against the histogram of the data.

But I think that it gradually became clear that the occasions when there was any practical advantage in fitting a curve to observed data were less than expected. Later, of course, in 1922, the criticism was raised by R. A. Fisher that the method of fitting by moments was often inefficient, although the practical significance of this inefficiency has still not been fully cleared up. But it so happened that other uses for these curves, other links with statistical theory, appeared over the horizon and that these were intimately connected with the development of statistical theory.

I am reminded of a remark which J. B. S. Haldene made about K.P. in a centenary address of 1957, in a rather different connection. Speaking of some of his work on heredity, Haldane said:

'I believe that this theory was incorrect in some fundamental respects. So was Columbus' theory of geography. He set out for China and discovered America! But he is not regarded as a failure for this reason.'

In the case of the frequency curves, it was not that they were incorrect but that the original purpose seemed to become less relevant to current statistical practice.

The first step in a new direction must undoubtedly be associated with the visit of W. S. Gosset ('Student') to K.P.'s Biometric Laboratory during two terms of the 1906-7 university session. At some date in the 1890's Arthur Guinness Son & Co. Ltd. initiated the policy of appointing to their staff in Dublin a number of university trained scientists. These young men found before them an almost unexplored field lying open for investigation. Because the idea was novel to industry, the firm did not want their rivals to be aware of what they had done, and for this reason all published work had to appear under a pseudonym. Hence the name 'Student'. Two later Guinness statisticians dubbed themselves 'Mathetes' and 'Sophister'.

Gosset was primarily a chemist, but having taken a year of Mathematics at Oxford, was put onto the interpretation of the Brewery's extensive records, both from the experimental brewery and from country-wide breeding experiments with barley. He had only at first available text-books on the theory of errors - Airy's and Merriman's, for example. In one direction he was searching for a measure of the relationship between two or more variables, having not come across the biometric work on correlation. Indeed, given a little more time, it seems likely that he would have re-discovered the correlation coefficient using an approach entirely different from Galton's. In another direction he was not happy about referring the ratio,  $(\bar{\mathbf{x}} - \mu)\sqrt{n}/s$ , to the normal probability scale when a sample contained only five or 10 observations, as had to be the case in much of the Brewery experimentation.

With K.P.'s guidance, which he acknowledged fully, he therefore carried out a programme of research involving the following steps.

- (a) Determining the 3rd and 4th moments of the sample estimate of variance,  $s^2$  , (the 1st and 2nd moments were already known).
- (b) Finding that the resulting moment ratios  $\beta_1(s^2)$ ,  $\beta_2(s^2)$  fell, for all sample sizes n , on the Type III, gamma or  $\chi^2$  line of the Pearson system.
- (c) Hence inferring that the distribution of s<sup>2</sup> was <u>likely</u> to be represented by a Type III distribution.
- (d) Proving that  $\bar{x}$  and  $s^2$  were uncorrelated in normal samples and hence, using (c), deriving the distribution of  $z = (\bar{x} \mu)/s$  (=  $t/\sqrt{n}$ ). This followed a Pearson Type VII curve.
- (e) Carrying out a random sampling experiment, with n=4, and so satisfying himself that the distributions which he had derived were likely to be correct. He took as his population a bivariate distribution of physical measurements on 3000 men and wrote these on small cards. Then he drew cards randomly in succession from a box, wrote the entries down in order and divided them consecutively into 750 samples of 4.

His famous paper 'On the probable error of a mean' appeared in <u>Biometrika</u> in 1908. In the same year he published another paper in which he used a similar intuitional process to infer that the distribution of the correlation coefficient, r, in samples of size n from uncorrelated normal material would be of the form

$$f(r) = const. (1 - r^2)^{\frac{n-4}{2}}$$

i.e., a Pearson Type II distribution. Again, he compared his theory with sampling results for n=4 and 8.

Now if we are concerned simply with priorities, we may note:

- (a) that in Germany, in the field of error theory, Abbe (1863) and Helmert (1876) had already derived the distribution of s<sup>2</sup> (although this work was quite unknown in England);
- (b) that Edgeworth in 1883, using prior probabilities, had derived a posterior distribution for the population mean,  $\mu$ , given  $\bar{x}$  and s, which was essentially Student's distribution.
- (c) that Edgeworth, in 1885, had also taken some samples from a rectangular distribution to show how quickly the mean tended to be normally distributed;
- (d) that Weldon had tossed dice in the 1890's to explore the nature and extent of sampling variation.

But this earlier work, apart perhaps from that of Weldon, was not closely related to the solution of practical problems and fell outside the main line of development of statistical theory. For 5-10 years little notice, either, was taken of Gosset's work, outside the Dublin Brewery. This I think was due to a number of reasons, which it is of interest to record briefly.

- (a) The early biometric work of Galton, Weldon and Pearson had not been concerned with laboratory experiments, but with collecting as large samples as possible 'in the field'. If you look through these early papers, as I have done recently, you will see that the analysis of data would rarely have gained from the use of any 'small sample' technique.
- (b) I suspect, also, that the early biometricians had a marked distrust of using small samples because of an instinctive feeling that much data were not homogeneous, either in space or time; that there would be small local variations which must be balanced out by collecting as large a sample as possible.

(c) Also there was a genuine fear that by advocating small-sample techniques, the biologist or medical man would think that he could draw conclusions, in fact unwarranted, from scanty data.

Of course, with hindsight we can criticize this attitude and say that small-sample theory was required to analyse such lack of homogeneity. But this concept was not yet born in the early 1900's.

However, Gosset had planted several seeds in a place where they were bound to grow, sooner or later. With regard to the history of frequency curves, he had shown how the Pearson system might be used to provide approximations (if not the true forms) to the sampling distributions of statistics whose moments only were known. This may be described as the second great use for these curves. He had also introduced the idea of making use of systematic random sampling to test, confirm or disprove mathematical theory.

Five or six years later, Student's work was followed up by R. A. Fisher who in 1912, a few months after graduating in Cambridge, derived the distributions of s<sup>2</sup>, and therefore of t, by an appeal to n-dimensioned space - a brilliant piece of imagination. Shortly afterwards he derived the sampling distribution of the correlation coefficient, r, for normal material and confirmed Student's guess about the distribution of r, when  $\rho = 0$ . All these results were published in <u>Biometrika</u> in 1915. By 1924 Fisher had shown that if  $\chi_1^2$  and  $\chi_2^2$  were two independent " $\chi^2$ 's", then  $\chi_1^2/\chi_2^2$  followed a Pearson Type VI and  $\chi_1^2/(\chi_1^2 + \chi_2^2)$  a Type I distribution. There was no difficulty, of course, in deriving these last two distributions, but no one had seen the need for them before Fisher had the simple but fundamental idea of comparing two independent estimates of variance, through their ratio. Student had in effect done this, but only in a particular case.

Thus entirely unexpectedly six of the seven main curves of K.P.'s system had been placed right at the centre of the newer statistical theory.

Pearson's system of curves had been first developed because so many parent distributions had been found <u>not</u> to be normal. What about the constant claim that the new tests alone were 'valid'? To some extent there were mathematical tools to probe this question of 'robustness'. The moments of  $\bar{x}$  and  $s^2$  were known in terms of population moments but not those of r, t, the variance ratio and so on.

Early in the 1920's the idea occurred of using random sampling experiments (Monte Carlo methods) to test these points, taking as the parent population not the normal distribution but some non-normal distribution of K.P.'s system. Here we had a third use for the curves. This work was encouraged by Gosset who, with his sound practical sense, was undoubtedly a little sceptical about claims of 'validity' without accompanying evidence. The problem of drawing random samples at first presented some difficulties. Gosset's 1906 use of slips of cardboard was laborious and it was difficult to ensure proper mixing. Church, working in the Biometric Laboratory at University College in 1924-25 at first used coloured glass beads, a different colour to represent each group in the population distribution. These could be easily mixed, but commercial coloured beads were not all of a size, and bias was introduced.

Then Tippett, who worked in the Laboratory in 1923-25, had the idea of using random numbers of which he made a standard table by selecting laboriously from census returns. He used these random numbers (1925) to check his approximations to the sampling moments of the extremes and the range in samples from a normal population. Apart from this particular application to which I shall refer again in a moment, the existence of this set of 10,000 four-figure random numbers, published in 1927, increased enormously the facility for examining the sensitivity to departure from normality of various 'normal theory' tests. Just as the introduction of

the Brunsviga calculator in 1894 had made an enormous simplification in statistical computation, so Tippett's <u>Numbers</u> provided the answer in the case of what we now term Monte Carlo investigations.

Besides the work of Church (1926) and Storey ('Sophister') (1928) I had a programme which involved taking random samples of 5, 10 and 20 from five or six non-normal populations, represented by Pearson curves. We were thus beginning to get some idea of the limits of the 'validity' claimed for Fisher's different tests; the 'robustness' varied considerably. This laborious work of the late 1920's would, of course, be very greatly speeded up now using a computer to generate pseudo-random numbers.

Turn back now for a moment to the distribution of the range,  $\omega$ . Gosset suggested it be used to provide quick estimates and tests of variability in place of the standard deviation. Using some rather more accurate estimates of the moments, which I had obtained in 1926, he assumed that we might represent the unknown true distribution of  $\omega$  by Pearson curves with the correct first four moments; so, in 1927, he published a table of approximate upper 10, 5 and 2 per cent points of  $\omega$  for n = 2(1)10 . Later (1932) I extended this work, providing four upper and four lower percent points for n = 2(1)30(5)100 . When accurate values were obtained in 1942 up to n = 20 , by direct computation, the corrections needed were very small. This was one of the most successful examples of how, what I have termed the second function of the frequency curves, could be turned to account long before the arrival of the computer age.

Other examples of this type of application were made at this time and because they were closely linked with some important theoretical developments, it is worth mentioning them here. In 1931 Jerzy Neyman and I published our paper termed 'On the problem of k samples' in a Polish journal.

Here we used the likelihood ratio principle to derive a test of homogeneity of variance. This was essentially what has since been termed 'Bartlett's test'. We found its sampling moments and approximated its distribution with a Type I (beta) curve. That same year I gave a summer session course in the University of Iowa where Sam Wilks was completing a thesis under H. L. Rietz. Part of this thesis, deriving some fundamental tests in multivariate analysis, for which I think he will have got some ideas from Harold Hotelling, was published in <u>Biometrika</u> in 1932. Wilks showed that his most important test criteria were likelihood ratio criteria, and gave their moments. In a joint paper of 1933, Sam and I took the case of two variables and applied the tests to some examples, using Type I curves to find the significance levels. Thus the curves provided a tool in developing tests derived from the likelihood ratio principle.

But there was another lead from the curves into statistical theory.

We were accumulating alternative tests regarding:

- (1) the population mean, i.e., those based on the sample mean, median and mid-range;
- (2) the population standard deviation, i.e., those based on the rootmean-square, the mean deviation and the range.

Also we had the possibility of using Tippett's <u>Random Numbers</u> to find their sampling distributions when the parent population was <u>not</u> normal. This almost inevitably suggested to us in the years 1929-31 the idea of studying the comparative power of alternative tests under various conditions. The theory of this method of approach was discussed by Neyman and myself in our paper of 1933. These examples could easily be extended, but I only want to indicate how these different lines of attack were interlinked.

Now let me come to a few last points. In his remarkable paper of

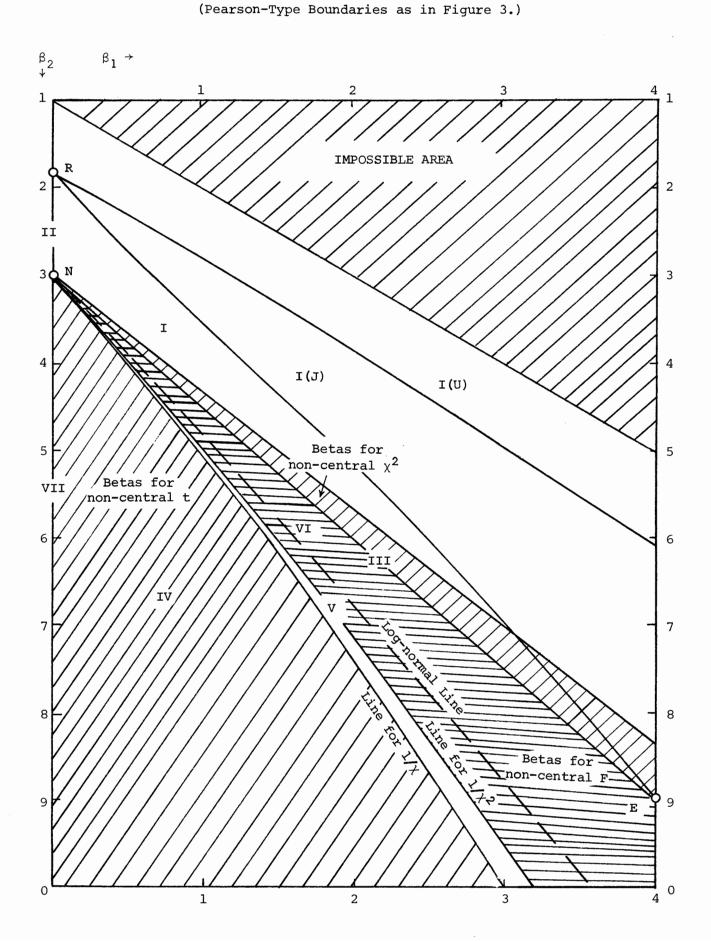
1928 Fisher had derived the distributions (in sampling normal material) of:

- (a) Multiple R (when  $\rho \neq 0$ ).
- (b) Non-central  $\chi^2 = \sum_{i=1}^{\nu} (x_i + a_i)^2$ .
- (c) Non-central F = non-central  $\chi^2(v_1)$ /central  $\chi^2(v_2)$ .
- (d) A little later Fisher (1931) had made use of the probability integral of non-central t =  $(X + \Delta)/\chi/\sqrt{\nu}$ , a function which Neyman was to make use of independently not long afterwards in Poland.

These multiparameter distributions could clearly assume a great variety of shapes, but did not (as their central counterparts) follow exactly distributions of the Pearson system. (b), (c) and (d) have been turned to use and tabled in connection with power functions. Tables of percentage points are also available for (b) and (d). But while it may be only of academic interest now, I have always wanted to relate the Pearson system to these non-central distributions if only as approximations. In the last few years it has become clear that we can, in fact, get surprisingly good approximations to (b), (c) and (d) by using a Pearson curve with the correct first four moments. The position is as shown in Figure 4. The beta points for non-central  $\chi^2$  lie in a narrow wedge-shaped region above the Type III or central  $\chi^2$  line; those for non-central F lie in the Type VI or inverted beta area, with a lower limit corresponding to the Type V or reciprocal of  $\chi^2$  line; while those for non-central t fall in the Type IV area, with an upper boundary on the line corresponding to the distributions of the reciprocal of  $\chi$  .

Finally with the cooperation of several others it was possible to issue (Johnson et al, 1963) a form of table which it had long seemed to me

FIGURE 4.  $\beta_1$  ,  $\beta_2$  REGIONS ASSOCIATED WITH NON-CENTRAL DISTRIBUTIONS.



would be valuable, namely a table of some 15 percentage points of Pearson curves, expressed in standard measure,

i.e. by 
$$X = (x - \mu)/\sigma$$
,

and entered with  $\sqrt{\beta_1}$  and  $\beta_2$  .

Apart from the fact that the table has already been used by several persons to derive approximate percent points of distributions for which the moments only are readily available, it acts as a kind of standard of comparison. We have been able to compare a number of distributions with those of the Pearson system, using the same first four moments, e.g., the lognormal (see Fig. 4), Johnson's  $S_{TI}$  and  $S_{R}$  curves (which transform into the normal and whose beta points lie, respectively below and above the lognormal line), the Weibull distribution (betas in Type I area), the non-central distributions I have just referred to and so on. There is generally pretty good agreement, even as far out as the 0.5% points, particularly at the long drawn out tail. Of course the distributions of these different systems will not have the same moments beyond the fourth, but such differences have their influence mainly in the extreme tails. One of the most striking things is agreement at the 5% points. As I have said, I am not sure of the practical importance of some of these results and relationships now, today, when we have the electronic computer. Nevertheless, the study has always fascinated me and, as an editor, I found on a number of occasions that my contributors would have made things clearer, indeed often have understood more the meaning of their results, had they had a greater familiarity with some of the long established properties of frequency curves.

All this has led me a long way from Galton's 'law of frequency of error' which the Greeks, had they known of it, would have deified! How

far statisticians have moved in the last 80 years is indeed well brought out by my final quotation from Galton's Natural Inheritance:

'It is difficult to understand why statisticians commonly limit their inquiries to Averages, and do not revel in more comprehensive views. Their souls seem as dull to the charm of variety as that of the native of one of our flat English counties, whose retrospect of Switzerland was that, if its mountains could be thrown into its lakes, two nuisances would be got rid of at once.'

(p. 63)

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13 ABSTRACT			

This paper is mainly concerned with the uses made of Karl Pearson's system of frequency curves, in particular with developments originating in the Department of Statistics at University College London. After a brief account of the history of the Normal curve from the time of DeMoivre to Francis Galton, Pearson's derivation of his system of curves from a single differential equation is described. The system was originally intended for the graduation of observed frequency distributions which the Normal curve would not fit. But since Student's pioneer work on the distributions of s², t and r (when  $\rho$  = 0) the curves have been used again and again to approximate the sampling distributions of statistics whose moments only are known. They have also been used in Monte Carlo experiments to examine the distribution of statistics based on samples taken from representative non-normal populations. The report illustrates some of this work and discusses conclusions to be drawn from it.

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