By means of the relation in (8) one deduces readily that:

$$(13) F(n,x) = \sum_{i=0}^{x} q_i^x / [(q_i - q_0) \cdots (q_i - q_{i-1})(q_i - q_{i+1}) \cdots (q_i - q_x)].$$

Jordan [1, p. 19, eq. (1)] shows this to be the xth Newton divided difference of  $q^n$  where the expansion is in terms of  $(q - q_0) \cdots (q - q_x)$ , for  $x = 0, 1, \cdots, n$ . The solution for (3) can now be written as:

(14) 
$$P(n, x) = (q - q_0) \cdots (q - q_{x-1})F_n(x)$$

from which follows:

$$\sum_{x=0}^{n} P(n,x) = q^{n}.$$

As remarked before, by setting q = 1 one obtains the solution of (1) subject to the boundary conditions (2).

It is clear that when all the  $q_i$  are equal that the Bernouilli distribution should come out as a special case. Since in this case the divided difference becomes the corresponding derivative divided by the appropriate factorial, one obtains:

(16) 
$$P(n, x) = \frac{(1 - q_0)^x}{x!} \frac{d^x q^n}{dq^x} \Big|_{q=q_0}.$$

Upon reduction this yields the usual formula, but not in the usual way.

By choosing  $p_x = \lambda_x/n$  and allowing n to increase without limit one obtains an analogue of the Poisson distribution, viz:

(17) 
$$P(x) = (-\lambda_0) \cdot \cdot \cdot (-\lambda_x) \sum_{i=0}^{x} e^{-\lambda_i} / [(\lambda_0 - \lambda_i) \cdot \cdot \cdot (\lambda_{i-1} - \lambda_i)(\lambda_{i+1} - \lambda_i) \cdot \cdot \cdot (\lambda_x - \lambda_i)]$$

which corresponds to the expansion of  $e^{-\lambda}$  about  $\lambda_0, \lambda_1, \lambda_2, \dots, \lambda_x, \dots$  when  $\lambda = 0$ .

## REFERENCE

 CHARLES JORDAN, Calculus of Finite Differences, Chelsea Publishing Co., New York, 2nd ed., 1947.

## A GRAPHICAL DETERMINATION OF SAMPLE SIZE FOR WILKS' TOLERANCE LIMITS

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1. Summary. To determine the smallest sample size for which the minimum and the maximum of a sample are the  $100\beta\%$  distribution-free tolerance limits at the probability level  $\epsilon$ , one has to solve the equation

(1) 
$$N\beta^{N-1} - (N-1)\beta^{N} = 1 - \epsilon$$

given by S. S. Wilks [1]. A direct numerical solution of (1) by trial requires rather laborious tabulations. An approximate formula for the solution has been indicated by H. Scheffé and J. W. Tukey [2], however an analytic proof for this approximation does not seem to be available. The present note describes a graph which makes it possible to solve (1) with sufficient accuracy for all practically useful values of  $\beta$  and  $\epsilon$ .

## 2. Construction of the graph. Substituting in (1)

$$N = \frac{\beta}{1 - \beta} x$$

we obtain

$$1 + x = (1 - \epsilon)\beta^{-\frac{\beta}{1-\beta}}x$$

and

(2) 
$$\log (1+x) = -\log \frac{1}{1-\epsilon} + \left(\frac{\beta}{1-\beta} \log \frac{1}{\beta}\right) x.$$

To solve (2) graphically, one has to find the intersection of the curve

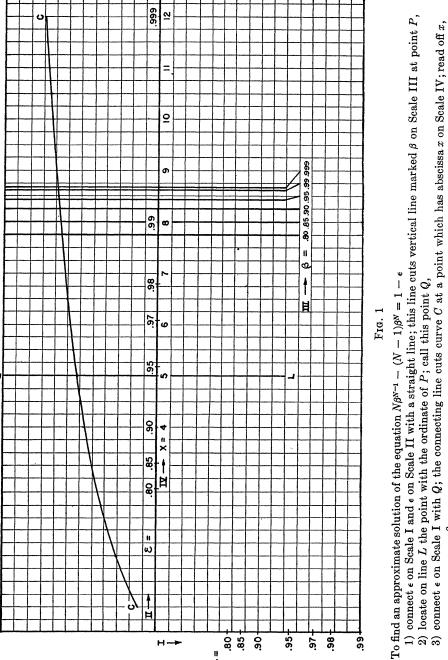
$$(3) y = \log(1+x)$$

with the line

$$y = -\log \frac{1}{1-\epsilon} + \left(\frac{\beta}{1-\beta}\log \frac{1}{\beta}\right)x.$$

To prepare a graph on which this can be done, one first plots (3) once for al (Figure 1, Curve C). Then one marks the points  $-\log\frac{1}{1-\epsilon}$  on the y-axis and labels them with the values of  $\epsilon$  (Figure 1, Scale I); chooses a constant r>0 and marks the points  $r\log\frac{1}{1-\epsilon}$  on the x-axis (Figure 1, Scale II); chooses a constant k>0, marks the points  $kr\frac{\beta}{1-\beta}\log\frac{1}{\beta}$  on the x-axis, draws vertical lines through each of these points, and labels them with the values of  $\beta$  (Figure 1, Scale III); draws the line x=k (Figure 1, line L); marks the uniform Scale IV on the x-axis.

The graph reproduced here has been prepared with r=4, k=5. It can easily be verified that the instructions on the graph lead to solutions x of (2) and  $N=x\frac{\beta}{1-\beta}$  of (1).



4) compute  $N = x \frac{\beta}{1 - \beta}$ 

3. Improvement by iterations. The graphical solution, usually accurate to two significant digits, may be improved easily by iterations. Replacing (2) by the equation

(4) 
$$x = \left[\log (1+x) + \log \frac{1}{1-\epsilon}\right] \left(\frac{\beta}{1-\beta} \log \frac{1}{\beta}\right)^{-1} = f(x)$$

one obtains iterations  $x_{j+1} = f(x_j)$  which, for .80  $\leq \epsilon \leq$  .999 and .80  $\leq \beta \leq$  .999 converge rapidly to the solution of (2).

Example. For  $\epsilon = .99$ ,  $\beta = .999$ , one finds graphically  $x_1 = 6.6$ , and from (4) the iteration formula  $x_{i+1} = \frac{\log (1 + x_i) + 2}{.4337}$  which yields the values  $x_2 = 6.642$ ,  $x_3 = 6.648$ ,  $x_4 = 6.649$ ,  $x_5 = 6.649$ . Rounding up we obtain the sample

size  $N = 6.649 \cdot 999 = 6643$ .

For  $\epsilon$  and  $\beta$  between .80 and .999 all iterations obtained from (4) are on the same side of the exact solution and converge to it monotonically. Thus, in our example, from  $x_1 < x_2$  we conclude that  $x_1$  as well as all further iterations are smaller than the exact solution.

## REFERENCES

- [1] S. S. Wilks, Mathematical Statistics, Princeton University Press, 1943, p. 94.
- [2] H. Scheffé and J. W. Tukey, "A formula for sample sizes for population tolerance limits," Annals of Math. Stat., Vol. 15 (1944), p. 217.