

Generation of Jute Fibre Length Distribution via Graphics and Computer Simulation with Gamma Distribution Function

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Abstract

Notwithstanding the complexity of production of jute fibre filaments a simple graphical presentation of jute fibre length distribution is made. The computer simulation of four types of frequency distribution of jute fibre length depending on method of sampling and testing is presented. These four distribution curves gives six points of intersections which merit technological significance. A study of fibre length-frequency distribution depending on methods of sampling and testing may lead to a general form of Gamma distribution function. The parameters in the mathematical formulae may serve as characteristics of condition giving the benchmark for valid experiment. This type of work involving exponential distribution was not done previously and provides a deeper insight into the characterisation of its comparative features with respect to normal distribution of fibre length. It can be observed that fibrograms for exponential and normal distributions have close similarity if their mean values are same. However, the survivor or array diagram distinguishes the basic distributions better than the fibrogram. This would be useful in the context that the basic frequency distributions of fibre lengths are somewhat difficult to obtain from direct experiments.

Keywords: Gamma distribution; Jute fibre; Fibre length; Length distribution

Introduction

In jute, the raw fibre as it comes from the plant consists of many single filaments bonded together by adhesion and branching to form a three-dimensional meshy structure. On carding, this meshy structure is broken into individual fibrous elements of different lengths, the longer elements being more complex, since these consists of several single filament still joined together in some way. This complexity will cause the linear density of the longer elements to be greater than that of the single filaments. In addition, the single filaments in the longer elements may be thicker and stronger than those in the shorter ones. On both these counts, the linear density of the longer elements will be greater than that of the shorter ones [1]. When the meshy structure of raw jute reeds is broken down randomly into fibrous elements in a carding machine, the frequency of such elements by number can be expected to decrease exponentially with increase in their length [2]. In this paper a graphical presentation of jute fibre length distribution is made. A study of fibre length-frequency distribution depending on methods of sampling and testing may lead to a general form of Gamma distribution [3] the density function of the random variable x is given by

$$f(x) = \left(\frac{1}{\Gamma(\alpha)} \right) \left(\lambda^\alpha x^{\alpha-1} e^{-\lambda x} \right), \alpha \geq 0$$

Whose

mean = α/λ , variance = α/λ^2 , standard deviation = $\sqrt{\alpha}/\lambda$ and coefficient of variation = $100/\sqrt{\alpha}$. This distribution has two parameters.

The parameter α is called a shape parameter since, as α increases, the density becomes more peaked. The parameter λ is a scale parameter; that is, the distribution depends on λ only. Variation for the values of α which may serve as behaviour characteristic of condition that appear as parameters in the mathematical formulae giving the benchmark for valid experiment.

Presentation of Fibre Length via Graphics

Figure 1 is the graphical representation of data collected by measuring length of single fibres. The vertical axis is number (quantity)

of fibre, the horizontal axis fibre length increments, and each circle o, represents a single fibre. One could imagine that the o's are end views fibres stacked vertically with each stack consisting only of fibres of same length. To the left hand side are the short fibres, the right hand side the long fibres. Statisticians call Figure 1 a histogram or length frequency distribution diagram. Figure 1 gives all of the information we need to know about fibre length and fibre length distribution. Since it is easier to determine the weight of fibre than to count the number of fibre, histograms and staple diagrams (arrays) are usually constructed on fibre weight basis. For clarity and ease of explanation, fibre number

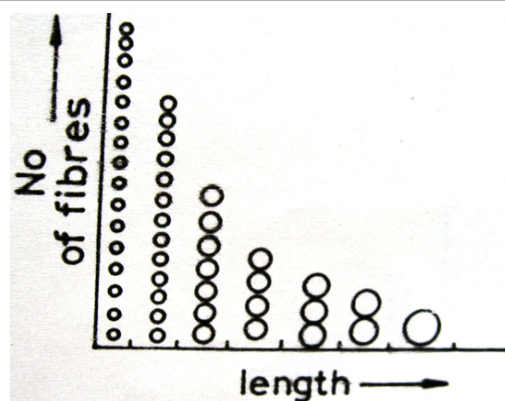


Figure 1: Histogram.

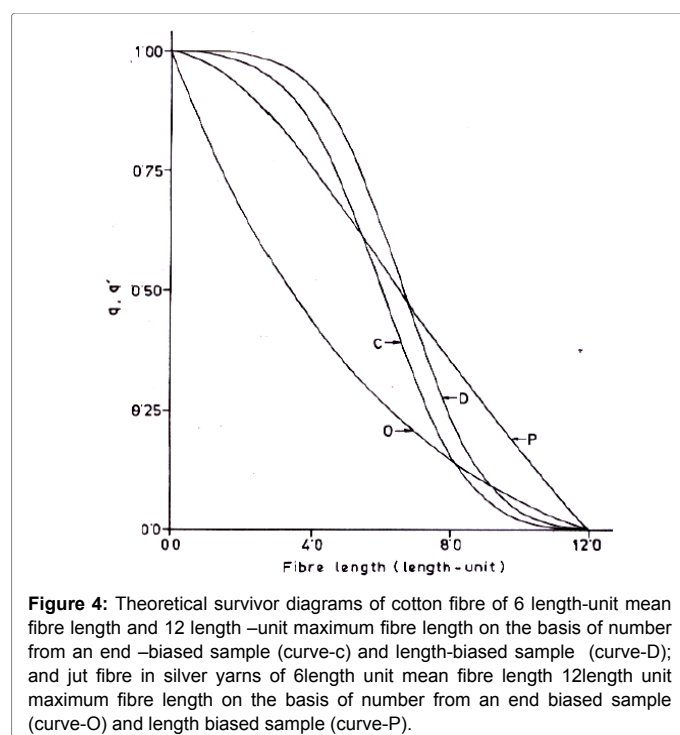
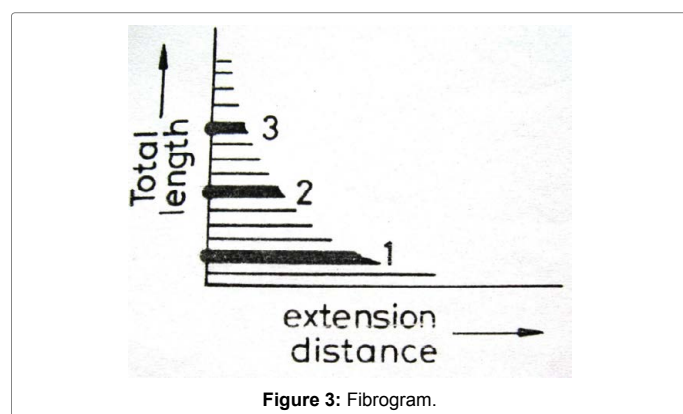
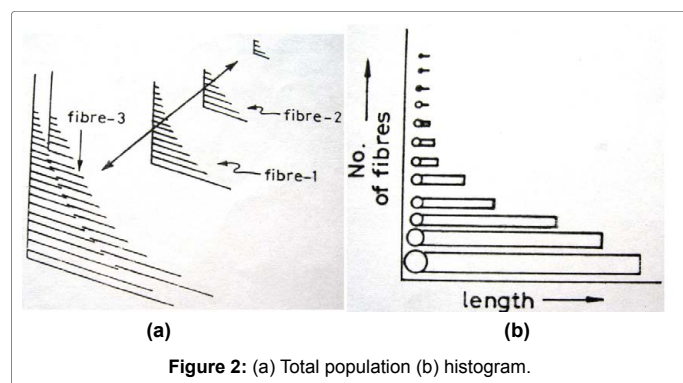
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basis is used in this discussion. Detailed discussion about number and weight basis of testing is made in later section of this paper. Another method of representing data collected by measuring length of fibres is in Figure 2. The vertical axis is number (quantity) of fibre, the horizontal axis is the fibre length and each cylinder represents a fibre. Figure 2a is a picture of fibres arranged in order of length with left hand ends aligned at the vertical axis. Figure 2a is simply a rearrangement of Figures 1 and 2 is the staple diagram and, as in Figure 1 contains all of the information we need to know about end to end lengths. Millions of millions of fibres are involved in fibre and yarn production. Figure 1 and Figure 2a do not contain all the fibres, they contain only a representative number of fibres. More realistic is Figure 2b which is three dimensional having a z-axis. Layer after layer of staple diagrams on to infinity. A bale of fibres could be represented as in Figure 2b. As fibres were rearranged going from Figure 1 to Figure 2b so can the fibres be rearranged as illustrated in Figure 3. To make this rearrangement, fibres of Figure 2b are selected at random along their lengths, the catch points placed on the vertical axis, and fibres thus selected arranged in descending order of the extension distance to the right hand side of the catch point. For illustration, fibre segments marked 1, 2, and 3 in Figure 2b are shown in their appropriate location in Figure 3 accordingly enough fibres are selected to make Figure 3 becomes representative. Figure 3 is a fibrogram and contains all of the information we need to know about fibre length and fibre length distribution. Of the three methods of representative of fibre length, the staple diagram of Figure 2a is easily accepted by our mental processes, the mental picture requires no strain on the imagination and we have a comfortable feeling; thus the staple diagram has been established as the standard of comparison and the bench mark from which other fibre length test methods are compared. However, in yarn spinning and non-oven fabric production, ends of fibres must never be aligned. The fibrogram which is an arrangement of fibre as they



are, or as they will be in future processes, becomes the bench mark of comparison concerning fibre performance in yarn spinning and non-oven fabric manufacture [4]. This type of work involving exponential distribution was not done previously and provides a deeper insight into the characterisation of its comparative features with respect to normal distribution of fibre length. The survivor diagrams (curves O and P) and fibrograms (curves Q and R) are also computed and presented in Figure 4 for end-biased and length-biased samples, respectively. It can be observed from Figure 4 that fibrograms for exponential and normal distributions have close similarity if their mean values are same. This is corroborated with the values of the corresponding fibrogram parameters. However the survivor or array diagram which is also still popular in industrial practice distinguishes the basic distributions as evident in Figure 5, better than the fibrogram. This would be useful in the context that the basic frequency distributions of fibre lengths are somewhat difficult to obtain from direct experiments.

Effect of sampling and testing method

In fibre length measurements two kinds of sample are experimentally obtainable through two kinds of sampling methods – end-biased and length biased. An end-biased sample is defined as one in which each fibre of population has equal probability of being included since each fibre has two ends independent of its length or any other characteristics. The end-biased sample is a random sample. A length biased sample is one in which the probability of a fibre being included is directly proportional to its length. In fibre testing, a random selection is restricted due to the essential nature of the fibre that it is much longer than it is thick. Because of this, it is only too easy to take a sample in such a way that it contains far more long fibres than it should. The knowledge of the nature of the bias in favour of the longer fibre would help to avoid the same. In some contexts in which length biased sampling is used it is reasonable to regard the length biased distribution as the object of study [5].

The frequency distributions of fibre length in jute slivers and yarns

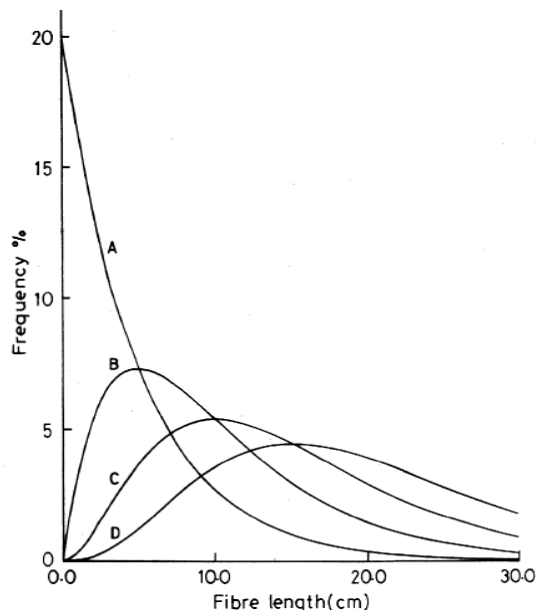


Figure 5: Theoretical Fibrograms of cotton fibre of 6 length-unit mean fibre length and 12 length-unit maximum fibre length on the basis of number from an end-biased sample (curve- E) and length-biased sample (curve-F); and jute fibre in sliver and yarns of 6 length-unit mean fibre length and 12 length-unit maximum fibre length on the basis of number from an end-biased sample (curve- Q) and length-biased sample (curve- R).

have been deduced analytically by Banerjee [2] for different methods of sample preparation. The expected average length of fibres obtained on the basis of number and weight of fibres has been estimated for each of such distribution. The present author [6-8] has shown that the Gamma function provides mathematical short-cut for the analysis of fibre length distribution and average length of jute fibres depending on sampling and testing method to arrive at the results as obtained by Banerjee. The two sampling method i.e., (i) end-biased and (ii) length biased sampling method have been considered. For each of such sampling the expected distribution of fibre length obtained on the basis of (i) number and (ii) weight of fibres has also been estimated. Thus four types of length distributions are obtained such as the frequency of length l tested on the basis of number from end-biased sample ($f_{EN}(l)$), the frequency of fibre of length l tested on the basis of number from a length-biased sample ($f_{LN}(l)$), the frequency of fibre of length l tested on the basis of weight from an end-biased sample ($f_{EW}(l)$) and the frequency of fibre of length l tested on the basis of weight from a length biased sample ($f_{LW}(l)$) (Figure 6).

The end-biased and number based length distribution of fibre $f_{EN}(l)$, which is expected to be similar to the basic type has been used by Banerjee (1980) to obtain $f_{LN}(l)$ and $f_{EW}(l)$, and in turn $f_{LN}(l)$ has been used to obtain $f_{LW}(l)$. The present author uses the Gamma function to obtain the same result very quickly [6-8] and $f_{EW}(l)$ has been considered [7] to obtain $f_{LW}(l)$. He has also considered all the other possible routes of transformations of one type of length distribution to the other types and presented their rederivation through Gamma function to achieve a considerable amount of simplification [9]. The summary of the results of the distribution is given in Table 1.

Cox [5] describes relations between mean (μ) of the end-biased distribution $f(x)$ and an expectation (mean or average) $E_g(x)$ with respect to length-biased distribution $g(x)$ as

$$E_g(X) = \mu \left(1 + \frac{\sigma^2}{\mu^2} \right)$$

Where μ, σ are the mean and standard deviation of the end-biased distribution $f(x)$.

For Gamma distribution the above relation becomes

$$E_g(X) = \frac{1}{\alpha} (1 + \alpha)$$

Where α is called a shape parameter of the end-biased distribution $f(x)$.

Computer simulation and modal length

The computer simulations of four frequency distribution function $f_{EN}(l)$, $f_{LN}(l)$, $f_{EW}(l)$ and $f_{LW}(l)$ as given in Table 1 were carried for $l=5$ cm. The computer programme in F77 is given in Appendix 1. The resulting curves obtained with the help of a Graphic Software are presented in Figure 6. However the proliferation of so many distribution curves depending on the method of sampling and testing may give several points of intersection when these are superimposed. In fact reports of experimental observation of intersection were available in some cases and attempt had been made to attribute suitable technological significance to it [8,10]. The present author has found out analytically all the intersections and showed that intersections depend on the average length of fibre population. Computer simulations are also presented and reported here [8].

He has also found out analytically the modal length of fibre length-frequency distribution depending on methods of sampling and testing [11,12]. The best measure of staple length for general application is given by modal or most frequent length of fibre length-frequency distribution [13].

The modal length l^m can be determined from the condition [14].

$$\text{at } x = l^m \quad (2)$$

$\frac{d}{dx} f(x) = 0$. Using Equations 1 and 2 can be written as

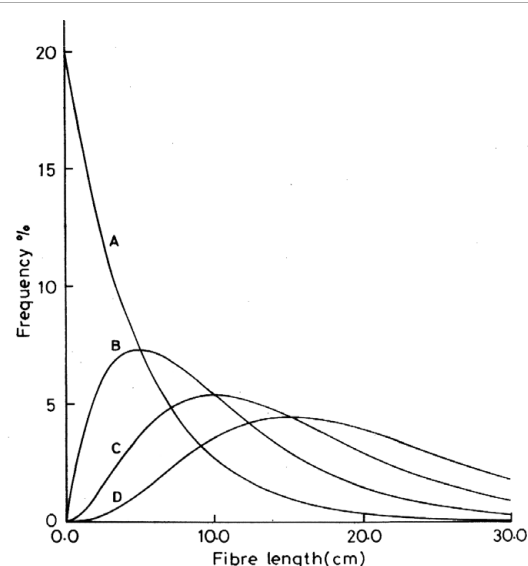


Figure 6: Theoretical length distribution of jute fibres in yarns and slivers. (A) End-biased number based method; (B) Length biased number based method; (C) End-biased weight based method; and (D) Length- biased weight based method).

Method followed	Relationship for frequency distribution	Average length
Basic population number-based	$f(I) = \frac{1}{I} \exp(-1/I)$	\bar{I}
End-biased sampling number-based	$f_{EN}(I) = \frac{1}{I} \exp(-1/I)$	\bar{I}
Length-biased sampling number-based	$f_{LN}(I) = \frac{1}{I^2} \exp(-1/I)$	$2\bar{I}$
End-biased sampling weight-based	$f_{EW}(I) = \frac{I^2}{2I^3} \exp(-1/I)$	$3\bar{I}$
Length-biased sampling weight-based	$f_{LW}(I) = \frac{I^3}{6I^4} \exp(-1/I)$	$4\bar{I}$

Table 1: Summary of results of derivation.

Method followed	Standard deviation	Coefficient of variation (%)
Gamma distribution (shape parameter= α , scale parameter $\lambda = 1/I$)	$\sqrt{\alpha}/\lambda$	$100/\sqrt{\alpha}$
Basic population Number-based Where $\alpha=1$	1	100
End-biased sampling Number –based Where $\alpha=1$	1	100
Length-biased sampling Number-based Where $\alpha=2$	$\sqrt{2}/1$	$100/\sqrt{2}=71$
End-biased sampling Weight-based Where $\alpha=3$	$\sqrt{3}/1$	$100/\sqrt{3}=58$
Length-biased sampling Weight-based Where $\alpha=4$	2/1	$100/2=50$

Table 2: Summary of results of standard deviation and coefficient of variation.

Method followed	Average length from Cox's formula	Modal length from equation (3)
Gamma distribution (shape parameter= α , scale parameter $\lambda = 1/I$)	$E_g(X) = \mu \left(1 + \frac{\sigma^2}{\mu^2} \right)$	$l^m = (\alpha - 1) / \lambda$
Basic population Number-based Where $\alpha=1$	Not required	0
End-biased sampling Number –based Where $\alpha=1$	Not required	0
Length-biased sampling Number-based Where $\alpha=2$	2/1	1
End-biased sampling Weight-based Where $\alpha=3$	3/1	2/1
Length-biased sampling Weight-based Where $\alpha=4$	4/1	3/1

Table 3: Summary of results of average length from cox's formula and modal length from equation (3).

$$\begin{aligned}
 & \left(\frac{1}{(\Gamma \alpha)} \right) \left(\lambda^\alpha \frac{d}{dx} (x^{\alpha-1} e^{-\lambda x}) \right) = 0 \\
 & \Rightarrow \left((\alpha-1)x^{\alpha-2} e^{-\lambda x} - \lambda x^{\alpha-1} e^{-\lambda x} \right) = 0 \\
 & \Rightarrow l^m = \frac{(\alpha-1)}{\lambda} \quad (3)
 \end{aligned}$$

Motivation and Conclusion

It may be mentioned that for Gamma distribution the density function of the random variable x is given by

$$f(x) = \left(\frac{1}{(\Gamma \alpha)} \right) \left(\lambda^\alpha x^{\alpha-1} e^{-\lambda x} \right), \alpha \geq 0$$

Whose

mean= α/λ , variance= α/λ^2 , standard deviation= $\sqrt{\alpha}/\lambda$ and coefficient of variation= $100/\sqrt{\alpha}$.

This distribution has two parameters. The parameter α is called a shape parameter since, as α increases, the density becomes more peaked. The parameter λ is a scale parameter; that is, the distribution depends on λ only. A study of fibre length-frequency distribution depending on methods of sampling and testing in Table 1 may lead to a general form of Gamma distribution function if we use $1/I=\lambda$. This remains the motivation of this work. Table 2 gives Standard Deviation and Coefficient of Variation for the respective values of α which may serves as characteristic of condition that appear as parameters in the mathematical formulae giving the benchmark for valid experiment [15,16]. It may be noted that not only average length of fibre but also standard deviation and coefficient variation depend on methods of sampling and testing. Table 3 gives Summary of Results of average length from Cox's formula and modal length from equation (3).

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