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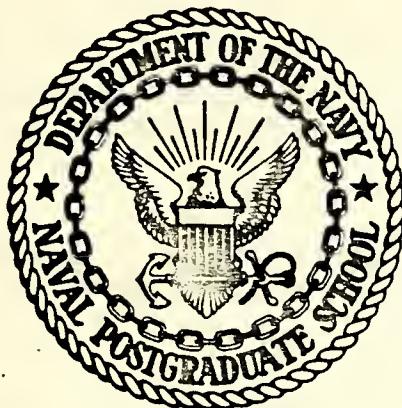
EMPIRICAL SAMPLING INVESTIGATION OF A GLOBAL  
MEASURE OF FIT OF PROBABILITY DENSITY  
FUNCTIONS

Li Hung Liu



# NAVAL POSTGRADUATE SCHOOL

## Monterey, California



# THESIS

EMPIRICAL SAMPLING INVESTIGATION OF A GLOBAL  
MEASURE OF FIT OF PROBABILITY DENSITY FUNCTIONS

by

Li Hung Liu

Advisor:

Peter A. W. Lewis

March 1974

Approved for public release; distribution unlimited.

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**Empirical Sampling Investigation of a Global  
Measure of Fit of Probability Density Functions**

**by**

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**Submitted in partial fulfillment of the  
requirements for the degree of**

**MASTER OF SCIENCE IN OPERATIONS RESEARCH**

**from the  
NAVAL POSTGRADUATE SCHOOL  
March 1974**



## **ABSTRACT**

The distribution of a measure of distance between a probability density function and its estimate is examined through Monte Carlo methods. The estimate of the density function is that proposed by Rosenblatt using sums of weight functions centered at the observed values of the random variables. The weight function used in all cases was triangular, but both uniform and Cauchy densities were tried for different sample sizes and bandwidths. The simulated distributions appear in all cases to be close to Gamma distributions, but it has not been possible to relate the parameters to the population characteristics or to the window shape and bandwidth.



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## I. INTRODUCTION

There are several recently proposed classes of empirical probability density function [Refs. 1-5] all generally considered to be superior to the classical histogram estimates. The class considered in this paper is based on independent observations, i.e.,  $x_1, x_2, \dots, x_n$  are independent and identically distributed random variables with continuous unknown density function  $f(x)$ . The method used to estimate  $f(x)$  is that proposed by Rosenblatt [4] and is as follows.

Denote the estimator of  $f(x)$  by  $f_n(x)$ . Then

$$f_n(x) = \frac{1}{nb(n)} \sum_{j=1}^n W\left\{\frac{x - x_j}{b(n)}\right\},$$

where  $W(u)$  is a bounded integrable weight function with

$$\int_{-\infty}^{\infty} W(u) du = 1$$

and  $b(n)$  is a bandwidth that tends to zero as  $n \rightarrow \infty$ , but is such that  $o(b(n)) = n^{-1}$ . i.e. it converges at a rate slower than  $1/n$ . Thus we might have  $b(n) \sim 1/n^{1/2}$ . Note that all estimates of this form for given observations are themselves density functions; that is

$$f_n(x) \geq 0,$$

$$\int_{-\infty}^{\infty} f_n(x) dx = 1.$$



Since the  $x_j$ 's are random variables,  $f_n(x)$  is a continuous parameter stochastic process.

The local properties of such estimates satisfying relatively mild conditions can be shown to be biased [3]. Our object in this thesis is to investigate a global measure of how good  $f_n(x)$  is as an estimate of  $f(x)$ . This measure was proposed by Bickel and Rosenblatt [2]. The particular global measure suggested for investigation is the statistic:

$$\beta(n) = \int \frac{[f_n(x) - f(x)]^2}{f(x)} dx ,$$

whose value will vary with each realization of  $x_1, \dots, x_n$ , i.e. it is a statistic or function of the n random variables.

Since an exact distribution of this statistic is not mathematically tractable, we have examined some representative cases by simulation.

The conjecture made by Bickel and Rosenblatt [2] is that the statistic  $\beta(n)$  is asymptotically normally distributed regardless of the underlying distribution  $f(x)$ . With this conjecture in mind, we hoped to establish a criterion that can be used to perform goodness of fit testing.

A problem with this type of criterion is that, unlike the Kolmogorov-Smirnov or Cramer-von-Mises statistics for goodness of fit of an empirical distribution function to the unknown distribution,  $\beta(n)$  is not distribution free and its distribution for finite n is unknown for any case. However, it is believed that  $\beta(n)$  is fairly robust with rapid convergence to the asymptotic distribution; another object of the thesis is to examine this hypothesis.



Going beyond asymptotic normality of  $\beta(n)$ , the following result for a quadratic functional is also of some interest [2]. The function  $a(x)$  is assumed to be a bounded piece-wise smooth integrable function. If  $b(n) = O(n^{-2/9})$  as  $n \rightarrow \infty$ ,

$$\{b(n)\}^{-1/2} [nb(n) \int [f_n(x) - f(x)]^2 a(x) dx - \int f(x) a(x) dx \int w(z)^2 dz]$$

is asymptotically normally distributed with mean zero and variance

$$2w^{(4)}(0) \int a(x)^2 f(x)^2 dx$$

as  $n \rightarrow \infty$ , where  $w^{(4)}$  is the 4-th convolution of  $w$  with itself.



## II. SIMULATION

The primary object of the simulation is to investigate the distribution of the statistic  $\beta(n)$ :

$$\beta(n) = \int \frac{[f_n(x) - f(x)]^2}{f(x)} dx ,$$

over a suitable range of integration. We have broken the simulation into the following cases:

### A. UNIFORM (0,1) CASE

Since

$$f(x) = \begin{cases} 1, & 0 \leq x \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

$\beta(n)$  becomes

$$\beta(n) = \int_{b(n)}^{1-b(n)} [f_n(x)-1]^2 dx .$$

We integrate from  $b(n)$  to  $1-b(n)$  instead of from 0 to 1, simply because we want to avoid the marked bias near 0 and 1. It can easily be shown that  $f_n(x)$  is biased in the end regions, but not otherwise.



Thus the mean and variance of  $f_n(x)$  in this case is:

$$\begin{aligned}
 E[f_n(x)] &= \frac{1}{b(n)} \int_0^1 W\left\{\frac{x-y}{b(n)}\right\} dy \\
 &= \frac{1}{b(n)} \int_{x-b(n)}^{x+b(n)} \left(1 - \frac{|x-y|}{b(n)}\right) dy \\
 &= \frac{1}{b(n)} \left[ \int_{x-b(n)}^{x+b(n)} dy - \int_{x-b(n)}^x \frac{(x-y)}{b(n)} dy - \int_x^{x+b(n)} \frac{(y-x)}{b(n)} dy \right] \\
 &= \frac{1}{b(n)} \left[ 2b(n) - \frac{1}{b(n)} \frac{(b(n))^2}{2} - \frac{1}{b(n)} \frac{(b(n))^2}{2} \right] = 1 .
 \end{aligned}$$

This is only true for  $b(n) \leq x \leq 1-b(n)$ . Also

$$\begin{aligned}
 \text{Var} &\left[ \frac{1}{nb(n)} \sum_{j=1}^n W\left\{\frac{x-X(j)}{b(n)}\right\} \right] \\
 &= \frac{1}{n^2 b^2(n)} \sum_{j=1}^n \text{Var} \left[ W\left\{\frac{x-X(j)}{b(n)}\right\} \right] \\
 &= \frac{1}{nb^2(n)} \text{Var} \left[ W\left\{\frac{x-X(j)}{b(n)}\right\} \right] \\
 &= \frac{1}{nb^2(n)} \left[ \int_0^1 W^2\left\{\frac{x-y}{b(n)}\right\} dy - \left( \int_0^1 W\left\{\frac{x-y}{b(n)}\right\} dy \right)^2 \right] .
 \end{aligned}$$

The integration was performed in the simulation using Simpson's rule, dividing the interval of integration into 100 equally spaced sub-intervals. The results were found to be satisfactory in the sense that the value of the integral for 100 sub-intervals differs little from those with more sub-intervals.



To evaluate  $f_n(x)$ , use the following formula:

$$f_n(x) = \frac{1}{nb(n)} \sum_{j=1}^n w\left\{\frac{x-x_{(j)}}{b(n)}\right\},$$

where the weight function  $w(u)$  is triangular, i.e.:

$$w(u) = \begin{cases} 1-|u|, & \text{if } |u| \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

$x_j$ ,  $j=1, \dots, n$  are independent and identically distributed random variables. In this first case, they are uniform (0,1) random variables. Also  $b(n)$  is the bandwidth. We employed three bandwidths in the uniform (0,1) case, namely:  $3/\sqrt{n}$ ,  $1/\sqrt{n}$ ,  $1/n$ , one at a time. For each bandwidth, we used five different sample sizes. i.e. 100, 200, 500, 1000, 1500, and for each sample size, we used five different random number seeds, i.e. simulated five times. The reason for this is to have the ability to assess the variability of the simulation results from the sections. The random number seeds are taken from the immediate previous "1500 sample size" output, except the first random number seed which is arbitrary. So, there are five sets of random numbers which are entirely different from one another. The following table should help to clarify the scheme:



TABLE I

Bandwidth	Sample size, n	No. replications	# Sections
$\frac{3}{n^{1/2}}$	100	2000	5
	200	2000	5
	500	2000	5
	1000	2000	5
	1500	2000	5
$\frac{1}{n^{1/2}}$	100	2000	5
	200	2000	5
	500	2000	5
	1000	2000	5
	1500	2000	5
$\frac{1}{n}$	100	2000	5
	200	2000	5
	500	2000	5
	1000	2000	5
	1500	2000	5

For each sample size, n, repeat five times using different random number seeds.

For the uniform (0,1) case, there are 75 outputs altogether. In each single simulation (one computer output), we evaluated 400  $\beta(n)$ 's, say  $\beta_j(n)$ ,  $j=1, \dots, 400$ . The output includes the following:

1. A bar histogram of the 400  $\beta(n)$ 's.
2. SNAP/IEDA graphs:
  - a. A refined solid histogram.
  - b. Graph of empirical log-survivor function,  $\ln(1-j/(N+1))$  vs  $\beta_{(j)}(n)$ , where  $\beta_{(j)}(n)$  is an ordered value.



c. Empirical CDF graph  $i/(n+1)$  vs  $\beta_{(j)}(n)$ , where  $\beta_{(j)}(n)$  denotes the ordered observed  $\beta_j(n)$ 's.

d. Normal probability distribution with estimated mean and variance. Norm vs e.c.d.f. of  $\beta_j(n)$ , where Norm is computed as follows: Let

$$\hat{\mu} = \frac{\sum \beta_j(n)}{n}$$

and  $s^2(\beta(n))$  = sample variance of the  $\beta_j(n)$ . Then

$$\text{Norm} = \Phi \left[ \frac{\beta_j(n) - \hat{\mu}}{s(\beta(n))} \right]$$

e. Normal probability plot Norm vs  $i/(N+1)$ .

A sample output is included in the "Computer Output" section (page 30 - 35).

For each bandwidth and each sample size, there are five computer outputs, or  $5 \times 400 = 2000$   $\beta_j(n)$ 's. With this 2000 sample size, we constructed a histogram again. Also we used the Rosenblatt smoothing technique to get a smooth density function estimate for  $\beta(n)$ . After a number of trials, we found that the bandwidth needed in the Rosenblatt estimator has very much to do with the smoothing and how "good" the density looks. Fortunately it is robust to a certain extent. This graph can give us a better picture of the true distribution than the classical histogram. The last graph of this complete output for sample size 2000 is the empirical log survival function;

$$\ln(1 - i/(N+1)) \text{ vs } \beta_{(j)}(n) .$$



There are 15 "2000 sample size" computer outputs in the uniform (0,1) case. (Since each bandwidth has five "2000 sample size" outputs.)

The complete computer outputs for each bandwidth and each sample size in the uniform (0,1) case are included in the "Computer Output" section (page 42 - 86).

## B. THE CAUCHY DISTRIBUTION

The next case investigated is the Cauchy distribution, again with a triangular weight function and with the range of integration (-3,3):

$$\beta(n) = \int_{-3}^{+3} \frac{[f_n(x) - f(x)]^2}{f(x)} dx .$$

The interval of integration comprises 80% of the probability mass for this distribution. The integration is carried out, in this case, using Simpson's rule with 600 sub-intervals. This grid was used after examining the 100, 300, 600, 900 sub-interval cases to ensure that a fine enough grid is used in the numerical integration.

The Cauchy density function is

$$f(x) = \frac{1}{\pi(1 + x^2)} .$$

The density function estimator is the same as in the uniform case:

$$f_n(x) = \frac{1}{nb(n)} \sum_{j=1}^n w\left\{\frac{x-x_j}{b(n)}\right\} .$$

The weight function  $w(u)$  is again triangular. Thus the only difference in this formula from the previous case is that  $x_j$ ,  $j=1, \dots, n$



are i.i.d.Cauchy random variables. Note also that for finite  $n$  the estimator will have a bias component, this bias component usually decreasing, for given  $n$ , as bandwidth decreases. In that case though the (pointwise) variance of  $f_n(x)$  increases and possibly also the variance of  $\beta(n)$ . Note that  $\beta(n)$  and its distribution is a function of both sample size and bandwidth.

The bandwidth  $b(n)$ 's used here are  $1/\sqrt{n}$ ,  $3/\sqrt{n}$ ,  $20/\sqrt{n}$ , the last one representing a case in which bias in the estimate  $f_n(x)$  plays a major role in the distribution of  $\beta(n)$ .

The rest of the simulation scheme is exactly the same as for the uniform  $(0,1)$  case.

A sample output for the 400  $\beta_j(n)$  case is included in the "Computer Output" section (page 36 - 41). Also the complete computer outputs (2000 replications) for each bandwidth and each sample size in the Cauchy case are included in the "Computer Output" section (page 87 -131).

#### C. HAND PLOT GRAPHS

Besides the computer work, there are six hand plot graphs. These are the  $\beta(n)$  quantiles versus the standard normal quantiles. They are included in the Appendix, to illustrate the rate of convergence of the distribution of  $\beta(n)$  to its asymptotic normal distribution.



### III. RESULTS

The asymptotic result, conjectured by Rosenblatt, is that by using a triangular weight function when estimating the uniform density function on  $(0,1)$

$$\{\beta(n)\}^{-1/2} \left\{ nb(n) \int_{b(n)}^{1-b(n)} |f(x)-1|^2 dx - (1-2b(n)) \int W(u)^2 du \right\}$$

is asymptotically normally distributed with mean zero and variance

$$2W^{(4)}(0)(1-2b(n))$$

as  $n \rightarrow \infty$  if  $nb(n) \rightarrow \infty$  and  $b(n) = o(n^{-2/9})$ . Here  $n$  is the sample size and  $b(n)$  the bandwidth. For the triangular weight function

$$\int_{-\infty}^{\infty} W(u)^2 du = \frac{2}{3}$$

and  $W^{(4)}(0)$  the 4th convolution of  $W$  with itself at zero, is  $302/630$ .

From the above expressions, we get:

$$E(\beta(n)) = E \left[ \int_{b(n)}^{1-b(n)} |f_n(x)-1|^2 dx \right] \sim \frac{2}{3} \frac{(1-2b(n))}{nb(n)} ,$$

$$\text{Var } (\beta(n)) = \text{Var} \left[ \int_{b(n)}^{1-b(n)} |f_n(x)-1|^2 dx \right] \sim \frac{2W^{(4)}(0)(1-2b(n))}{n^2 b(n)} .$$

A comparison of the simulated values with that of the conjectured is tabulated below for the different cases investigated when  $f(x)$  is a uniform density:



TABLE II

Comparison of estimated mean values and asymptotic mean values of  $\beta(n)$  for different bandwidths and sample sizes.

n	$b(n) = 3/\sqrt{n}$	$E(\beta(n))$	$E(\beta(n))/(1-2b(n))$	
			Conjectured	Computer output
100	.3000	.0089	.0222	.0127
200	.2121	.0090	.0157	.0109
500	.1342	.0073	.0099	.0075
1000	.0949	.0057	.0070	.0058
1500	.0775	.0048	.0057	.0051

n	$b(n) = 1/\sqrt{n}$	$E(\beta(n))$	$E(\beta(n))/(1-2b(n))$	
			Conjectured	Computer output
100	.1000	.0533	.0667	.0583
200	.0707	.0405	.0471	.0415
500	.0447	.0271	.0298	.0269
1000	.0316	.0197	.0211	.0197
1500	.0258	.0163	.0172	.0168

TABLE III

Comparison of estimated standard deviation values and asymptotic standard deviation values of  $\beta(n)$  for different bandwidths and sample sizes.

n	$b(n) = 3/\sqrt{n}$	$\sigma(\beta(n))$	$\sigma(\beta(n))/(1-2b(n))$	
			Conjectured	Computer output
100	.3000	.0113	.0283	.0115
200	.2121	.0081	.0141	.0088
500	.1342	.0046	.0063	.0047
1000	.0949	.0029	.0036	.0030
1500	.0775	.0022	.0026	.0023

n	$b(n) = 1/\sqrt{n}$	$\sigma(\beta(n))$	$\sigma(\beta(n))/(1-2b(n))$	
			Conjectured	Computer output
100	.1000	.0277	.0346	.0315
200	.0707	.0171	.0199	.0189
500	.0447	.0088	.0097	.0092
1000	.0316	.0053	.0057	.0056
1500	.0258	.0040	.0042	.0043



Especially for small bandwidth the agreement between the asymptotic and simulated variances is very good even for small n(n=100). The same is true for the expected value, although convergence is slower than for  $\sigma$  and again slower for large bandwidth.

The simulation reveals that the distribution of the  $\beta(n)$  statistic converges to normality very slowly (cf. the six quantile plots in the Appendix). For sample sizes 100, 200, 500, 1000, 1500, the histograms are all skewed to the right; the empirical log-survivor function plots are concave downward; and none of the normal probability plots is approximately straight lines (refer to "Computer Output" section). It is very suggestive that the  $\beta(n)$  statistic is approximately Gamma ( $\theta, k$ ) distributed where the density of the Gamma distribution is:

$$f(x) = \frac{(k/\theta)^k e^{-kx/\theta} x^{k-1}}{(k-1)!}$$

and the mean and variance are:

$$E(X) = \theta ; \quad \text{Var}(X) = k\theta^2 .$$

Therefore, for each underlying distribution, each bandwidth, and each sample size, we used the 2000 evaluated  $\beta(n)$ 's to estimate the  $\theta$ 's and  $k$ 's, and tabulate them as follows:



TABLE IV

Estimated parameters for fitted Gamma distribution for  $\beta(n)$  in the case where the densities are uniform (0,1) and Cauchy, with a triangular weight function.

- (1) Bandwidth =  $1/n$  Underlying distribution: Uniform (0,1)

Sample sizes and estimated  $\theta$ 's and k's

	100	200	500	1000	1500 (std. dev.)
$\tilde{\theta}$	.6540	.6638	.6612	.6621	.6645 $\pm$ .0045
$\tilde{k}$	39.48	38.46	32.95	30.86	31.06 $\pm$ 2.91

- (2) Bandwidth =  $1/\sqrt{n}$  Underlying distribution: Uniform (0,1)

Sample sizes and estimated  $\theta$ 's and k's

	100	200	500	1000	1500 (std. dev.)
$\tilde{\theta}$	.0565	.0420	.0279	.0201	.0166 $\pm$ .0003
$\tilde{k}$	3.62	5.56	8.63	12.61	16.92 $\pm$ 1.45

Underlying distribution: Cauchy

Sample sizes and estimated  $\theta$ 's and k's

	100	200	500	1000	1500 (std. dev.)
$\tilde{\theta}$	.3921	.2800	.1764	.1258	.1029 $\pm$ .0004
$\tilde{k}$	21.02	31.49	59.34	79.19	100.83 $\pm$ 4.37

- (3) Bandwidth =  $3/\sqrt{n}$  Underlying distribution: Uniform (0,1)

Sample sizes and estimated  $\theta$ 's and k's

	100	200	500	1000	1500 (std. dev.)
$\tilde{\theta}$	.0123	.0107	.0079	.0060	.0051 $\pm$ .0001
$\tilde{k}$	1.11	1.59	2.43	3.65	5.03 $\pm$ 0.37

Underlying distribution: Cauchy

Sample sizes and estimated  $\theta$ 's and k's

	100	200	500	1000	1500 (std. dev.)
$\tilde{\theta}$	.1247	.0911	.0576	.0412	.0340 $\pm$ .0004
$\tilde{k}$	8.62	12.30	20.32	29.14	33.98 $\pm$ 1.55

- (4) Bandwidth =  $20/\sqrt{n}$  Underlying distribution: Cauchy

Sample sizes and estimated  $\theta$ 's and k's

	100 (std.dev.)	200	500	1000	1500 (std. dev.)
$\tilde{\theta}$	.0558 $\pm$ .0008	.0262	.0110	.0066	.0052 $\pm$ .0001
$\tilde{k}$	6.71 $\pm$ 0.78	4.03	3.21	4.10	5.24 $\pm$ 0.16

where  $\tilde{\theta}$  = mean ,

$$\tilde{k} \approx 1/(\text{estimated coefficient of variation})^2$$



The standard deviation estimates for the parameter estimates are obtained from the five identical simulations using different random number seeds, each generating a sample size of 400, as detailed in **Table I.**



#### **IV. CONCLUSIONS**

Although the conjecture proposed by Rosenblatt is that the distribution of the statistic  $\beta(n)$  is asymptotically normal, this is certainly not the case for moderate sample sizes such as those used in the simulations. Based on this fact, the conjecture has little practical value, but the fact that all cases examined give rise to an approximate Gamma distribution gives some hope for stability in the results.

However, the method proposed by Rosenblatt for estimating density functions is no doubt plausible as can be seen from the Rosenblatt type estimates used in this study. More effort in this area is needed, particularly to relate the Gamma distribution parameters to underlying factors. Since we realize that for moderate sample sizes, the statistic  $\beta(n)$  is not distribution free, and the bandwidth  $b(n)$  (a function of sample size) plays a crucial role in estimating the density function and also the shape of the weight function  $W(u)$  is a major factor for evaluating  $f_n(x)$ , the future work may be to find the best combination of the above three factors by simulating the distribution of the statistic  $\beta(n)$  for different underlying distributions, bandwidths, sample sizes and weight functions and tabulate the resulting quantiles. Then we may be able to put the statistic  $\beta(n)$  to actual use.



## APPENDIX

Included in this Appendix are six plots of sample size versus estimated quantiles of the statistic  $\beta(n)$ , normalized by subtracting the estimated mean and dividing by the estimated standard deviation. Also plotted as straight lines are the corresponding quantiles of the unit normal distribution.

The 17 quantiles, extracted from each 2000- $\beta(n)$  sample, are 0.010, 0.020, 0.025, 0.05, 0.10, 0.20, 0.30, 0.40, 0.50, 0.60, 0.70, 0.80, 0.90, 0.95, 0.975, 0.980, 0.990.

Other data for each plot are self-explanatory.



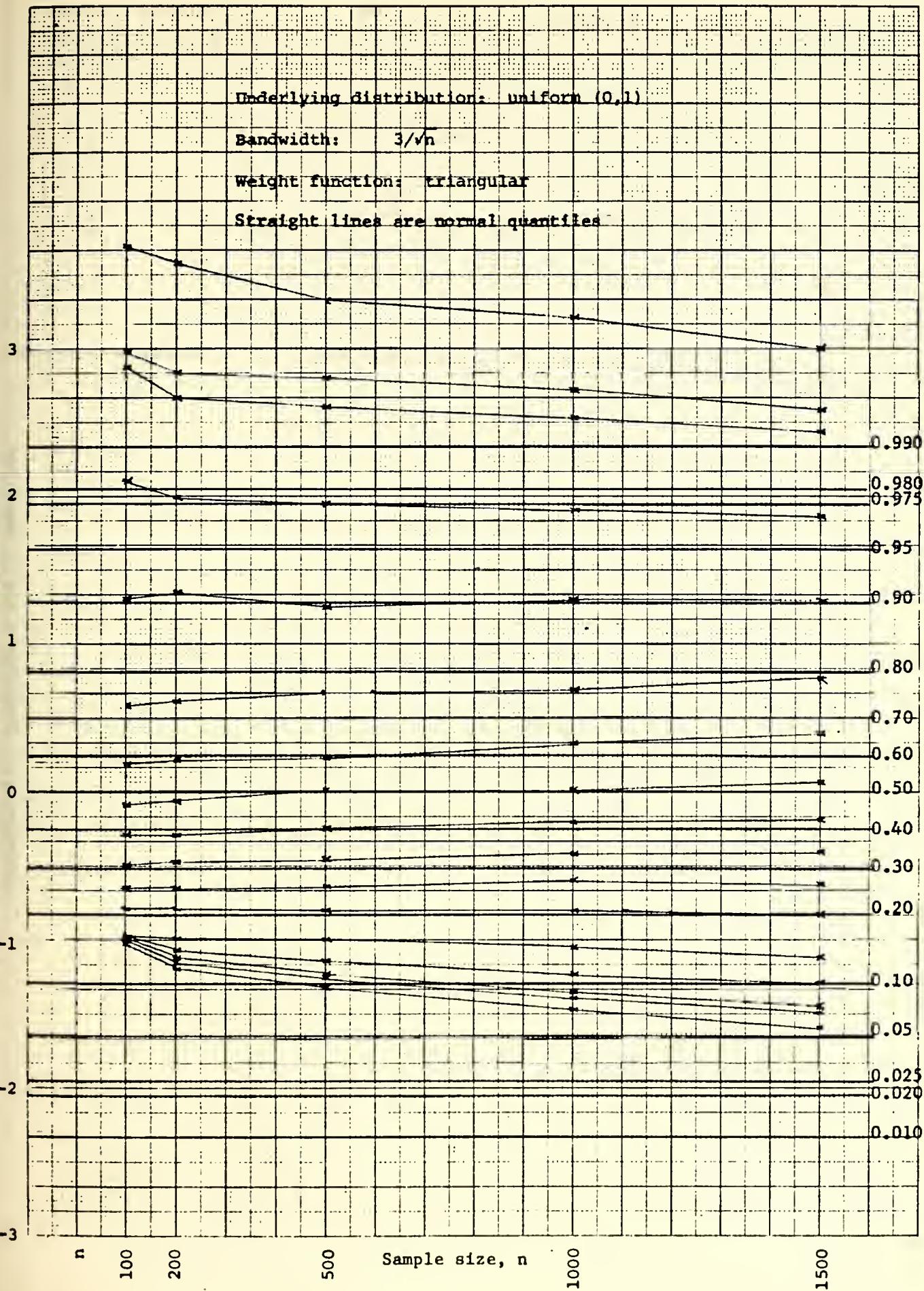
Underlying distribution: uniform (0,1)

Bandwidth:  $3/\sqrt{n}$

Weight function: triangular

Straight lines are normal quantiles

Estimated  $\beta(n)$  quantiles &  $N(0,1)$  quantiles



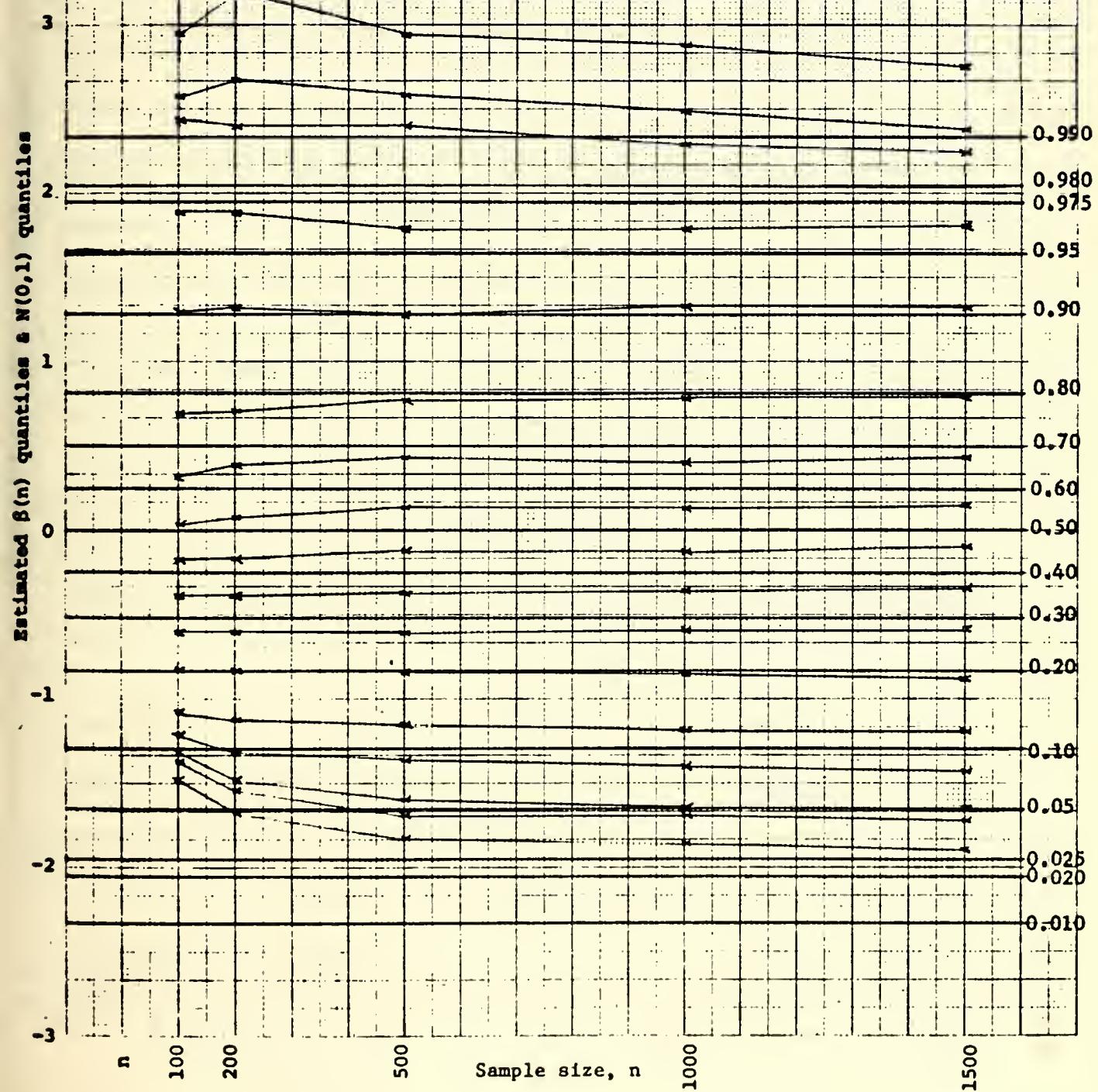


Underlying distribution: uniform (0,1)

Bandwidth:  $1/\sqrt{n}$

Weight function: triangular

Straight lines are normal quantiles



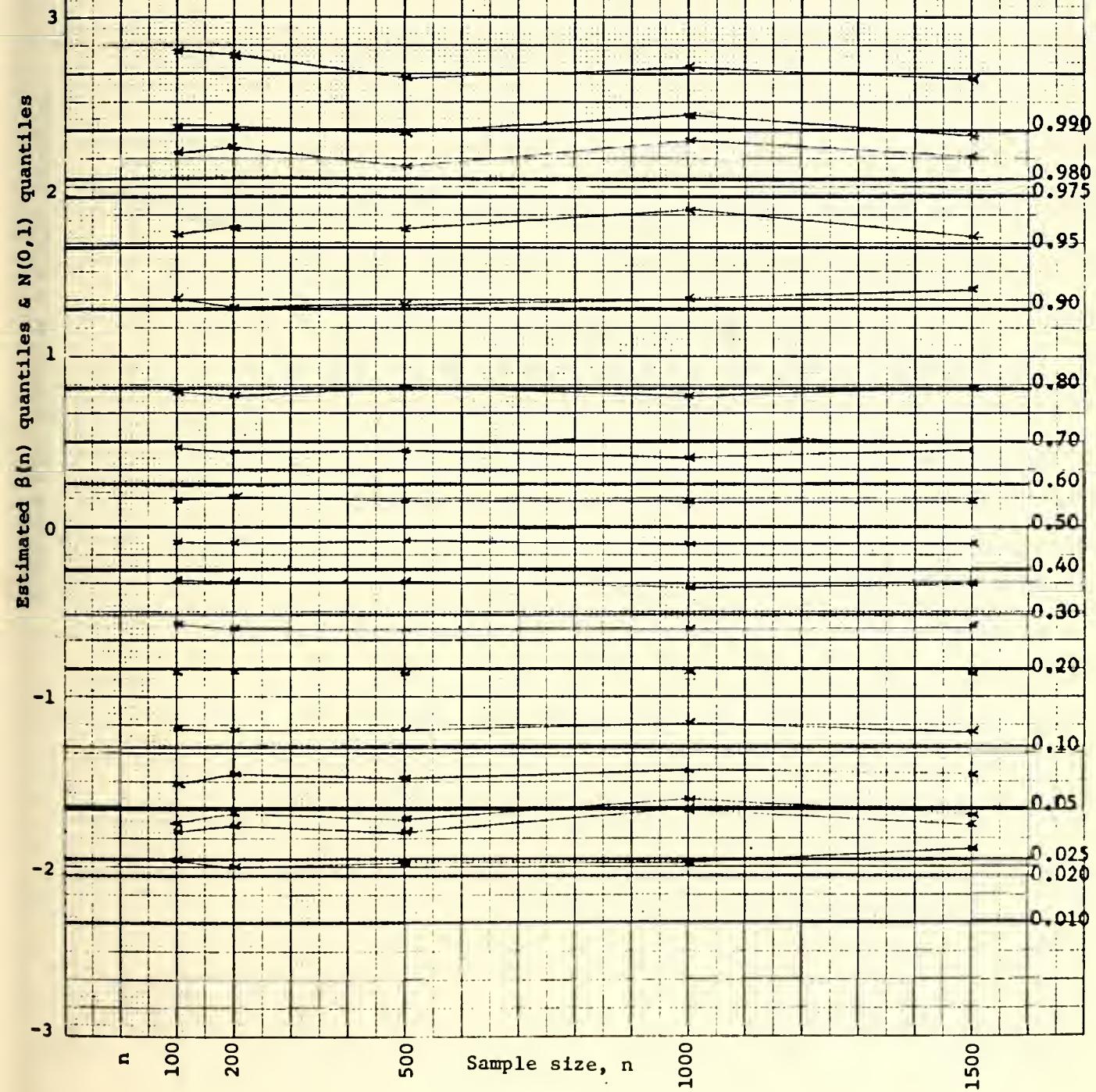


Underlying distribution: uniform (0,1)

Bandwidth:  $1/n$

Weight function: triangular

Straight lines are normal quantiles.



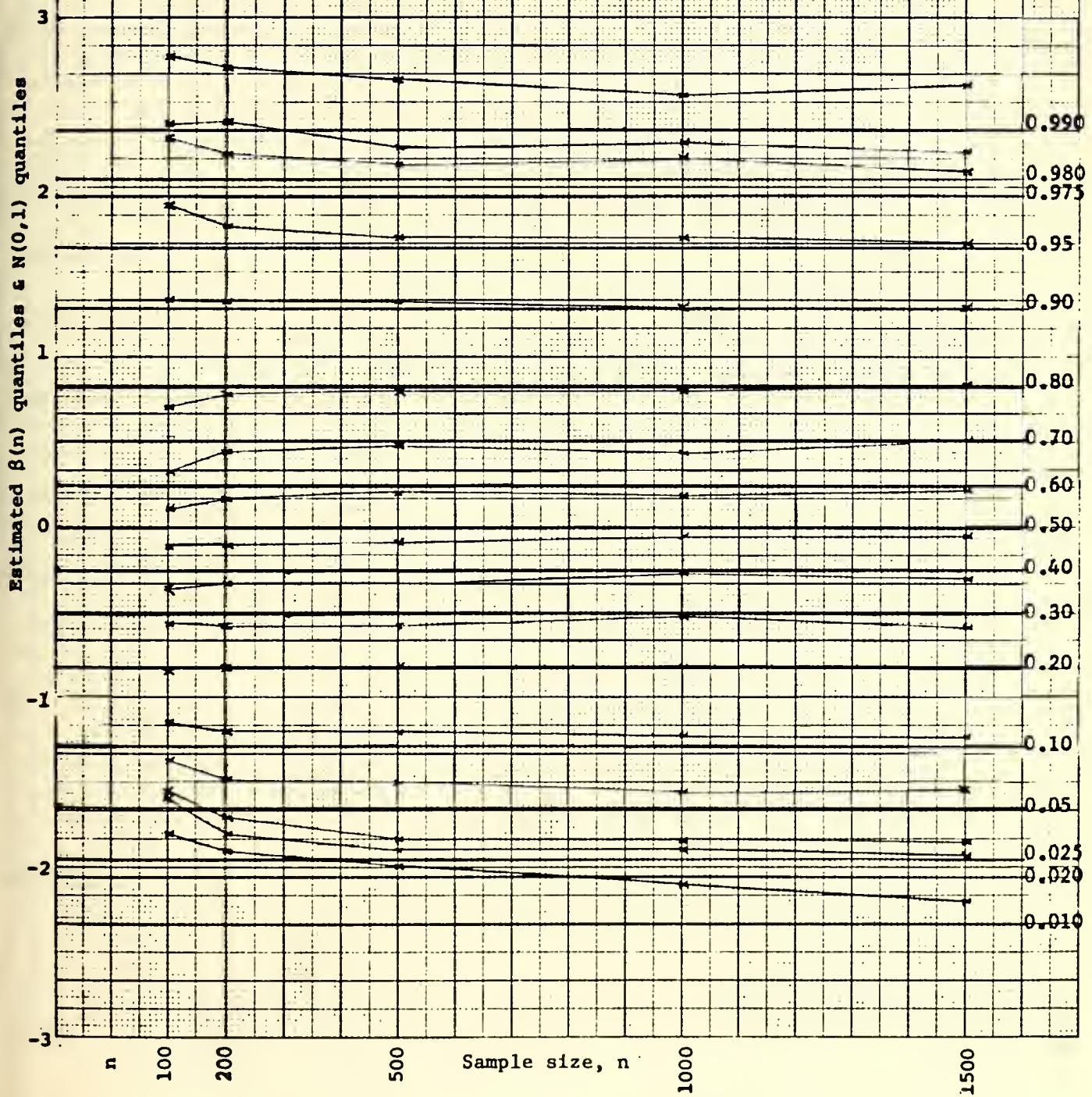


Underlying distribution: Cauchy

Bandwidth:  $1/\sqrt{n}$

Weight function: triangular

Straight lines are normal quantiles



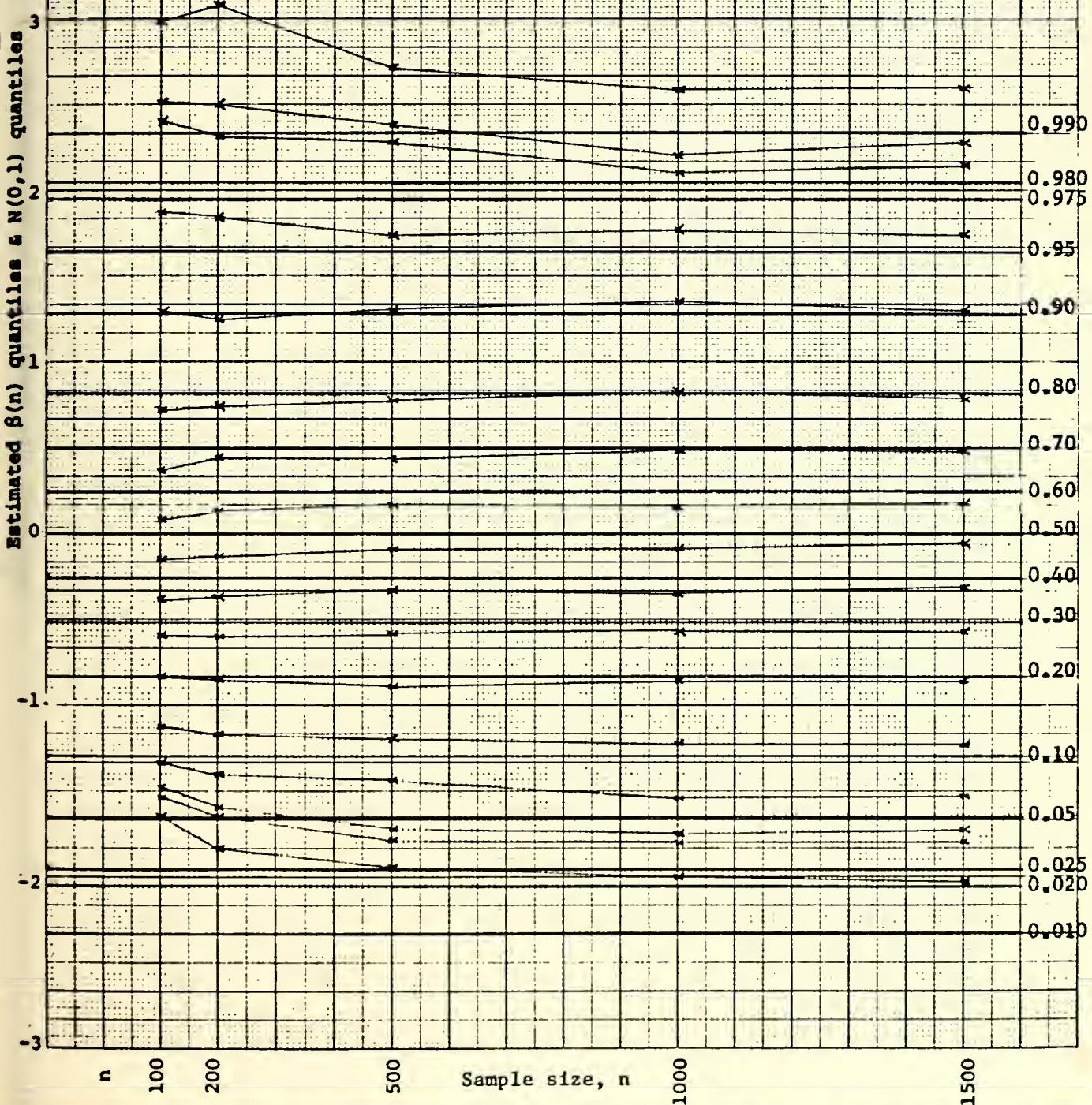


Underlying distribution: Cauchy

Bandwidth:  $3/\sqrt{n}$

Weight function: triangular

Straight lines are normal quantiles





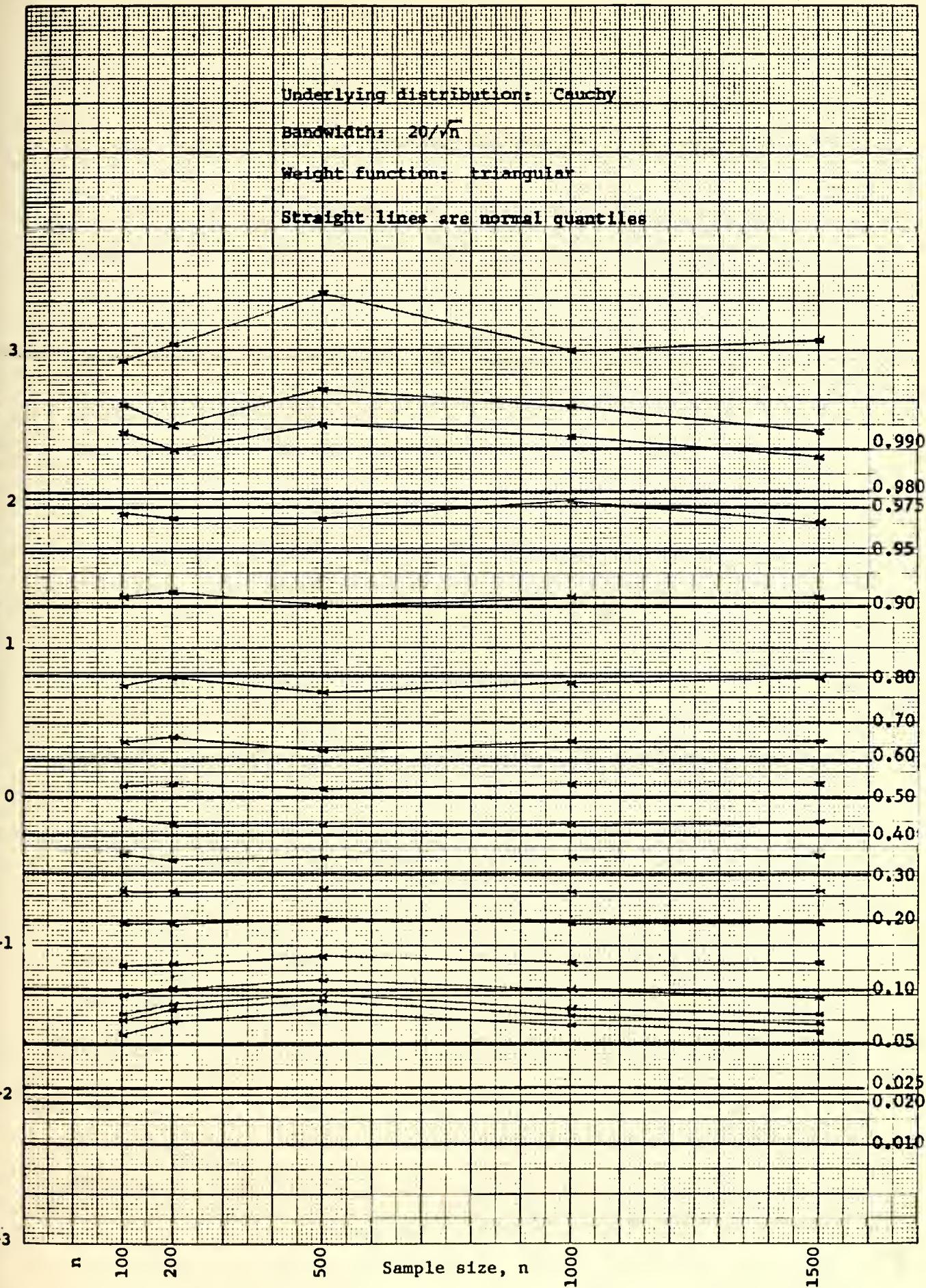
Underlying distribution: Cauchy

Bandwidth:  $20/\sqrt{n}$

Weight function: triangular

Straight lines are normal quantiles

Estimated  $\beta(n)$  quantiles &  $N(0,1)$  quantiles





## COMPUTER OUTPUT

There are altogether 102 sheets of graphs. All of them are self-explanatory, except that the letter "X" in the graphs represents the statistic  $\beta(n)$ .

The following is a list of index of the graphs.

<u>Page</u>	<u>Description</u>
30 - 35	sample computer output for the case: underlying distribution-uniform (0,1); underlying distribution sample size n=1500; bandwidth $3/\sqrt{n}$ ; statistic $\beta(n)$ sample size m=400.
36 - 41	sample computer output for the case: underlying distribution-Cauchy; underlying distribution sample size n=100; bandwidth $1/\sqrt{n}$ ; statistic $\beta(n)$ sample size m=400.

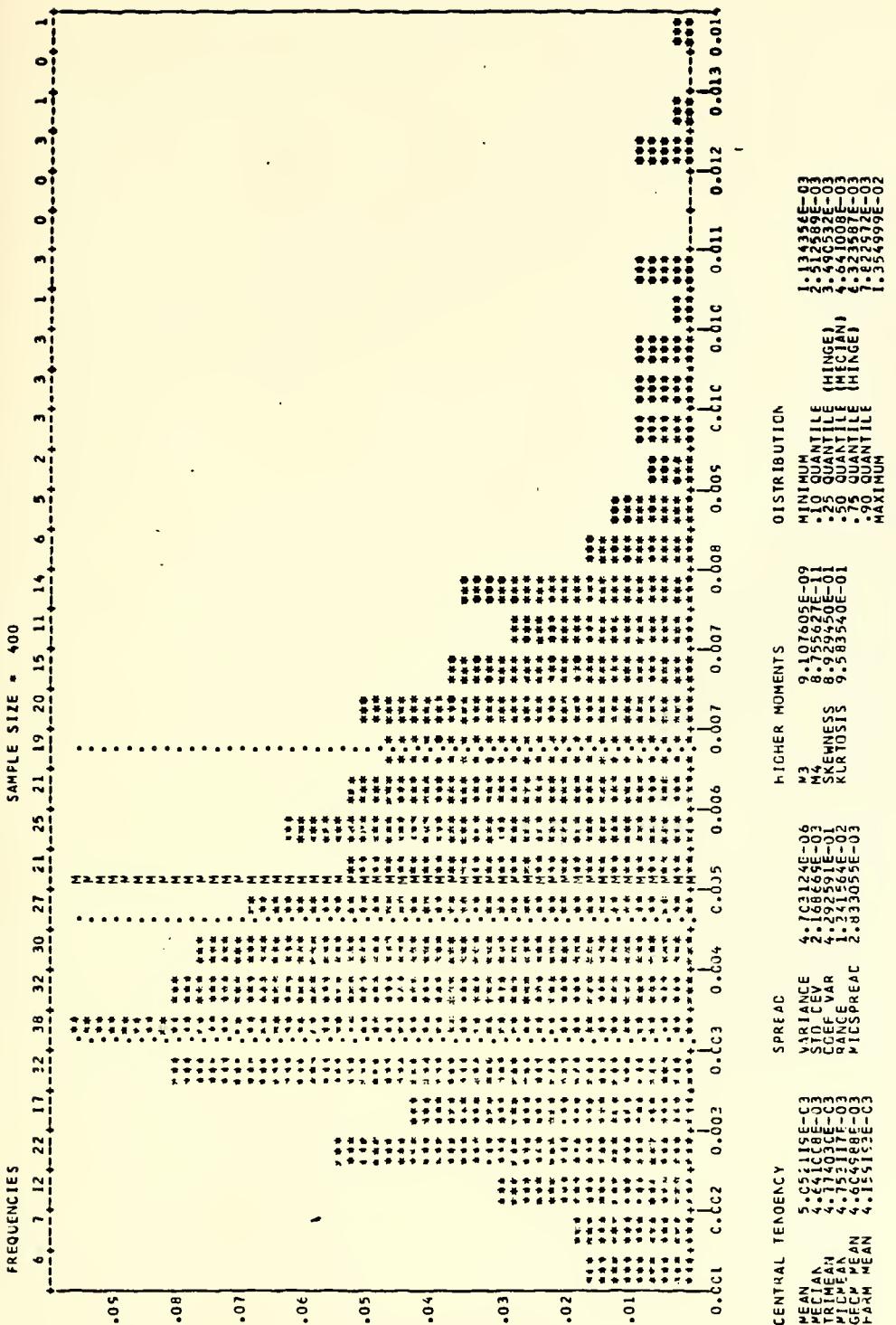
The rest of the computer graphs in this section are all for statistic  $\beta(n)$  sample size m=2000. The page index is as follows.

<u>Page</u>	<u>Underlying distribution</u>	<u>Underlying distribution sample size</u>	<u>Bandwidth</u>
42 - 44	U(0,1)	100	$3/\sqrt{n}$
45 - 47	U(0,1)	200	$3/\sqrt{n}$
48 - 50	U(0,1)	500	$3/\sqrt{n}$
51 - 53	U(0,1)	1000	$3/\sqrt{n}$
54 - 56	U(0,1)	1500	$3/\sqrt{n}$
57 - 59	U(0,1)	100	$1/\sqrt{n}$
60 - 62	U(0,1)	200	$1/\sqrt{n}$



Page	Underlying distribution	Underlying distribution sample size	Bandwidth
63 - 65	U(0,1)	500	$1/\sqrt{n}$
66 - 68	U(0,1)	1000	$1/\sqrt{n}$
69 - 71	U(0,1)	1500	$1/\sqrt{n}$
72 - 74	U(0,1)	100	$1/n$
75 - 77	U(0,1)	200	$1/n$
78 - 80	U(0,1)	500	$1/n$
81 - 83	U(0,1)	1000	$1/n$
84 - 86	U(0,1)	1500	$1/n$
87 - 89	Cauchy	100	$1/\sqrt{n}$
90 - 92	Cauchy	200	$1/\sqrt{n}$
93 - 95	Cauchy	500	$1/\sqrt{n}$
96 - 98	Cauchy	1000	$1/\sqrt{n}$
99 - 101	Cauchy	1500	$1/\sqrt{n}$
102- 104	Cauchy	100	$3/\sqrt{n}$
105- 107	Cauchy	200	$3/\sqrt{n}$
108- 110	Cauchy	500	$3/\sqrt{n}$
111- 113	Cauchy	1000	$3/\sqrt{n}$
114- 116	Cauchy	1500	$3/\sqrt{n}$
117- 119	Cauchy	100	$20/\sqrt{n}$
120- 122	Cauchy	200	$20/\sqrt{n}$
123- 125	Cauchy	500	$20/\sqrt{n}$
126- 128	Cauchy	1000	$20/\sqrt{n}$
129- 131	Cauchy	1500	$20/\sqrt{n}$





30



c. tcc

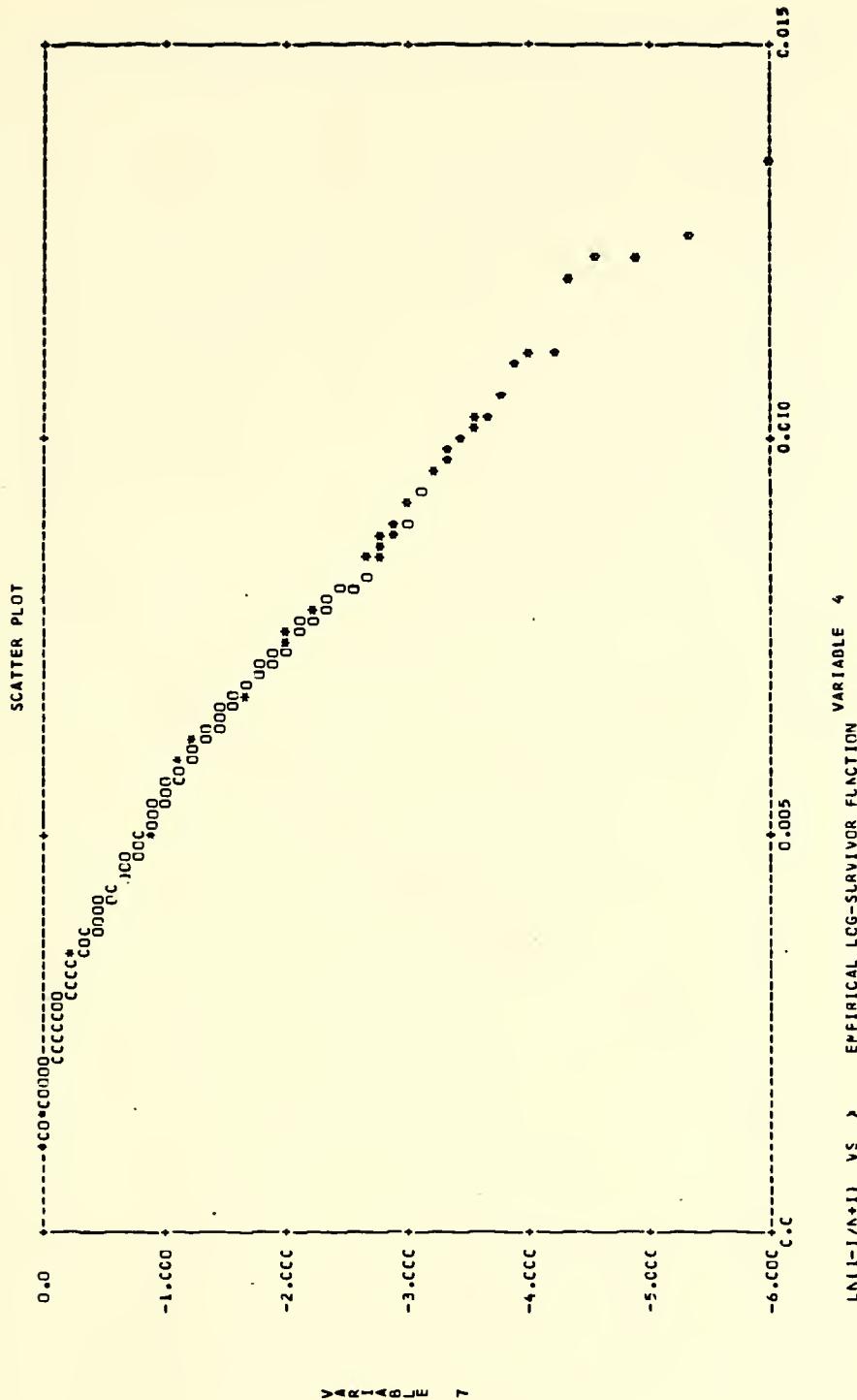
0.100

c. ccc

0.0001 0.002 0.004 0.005 0.006 0.007 0.008 0.010 0.011 0.012 0.013 0.013

VARIABLE . 4



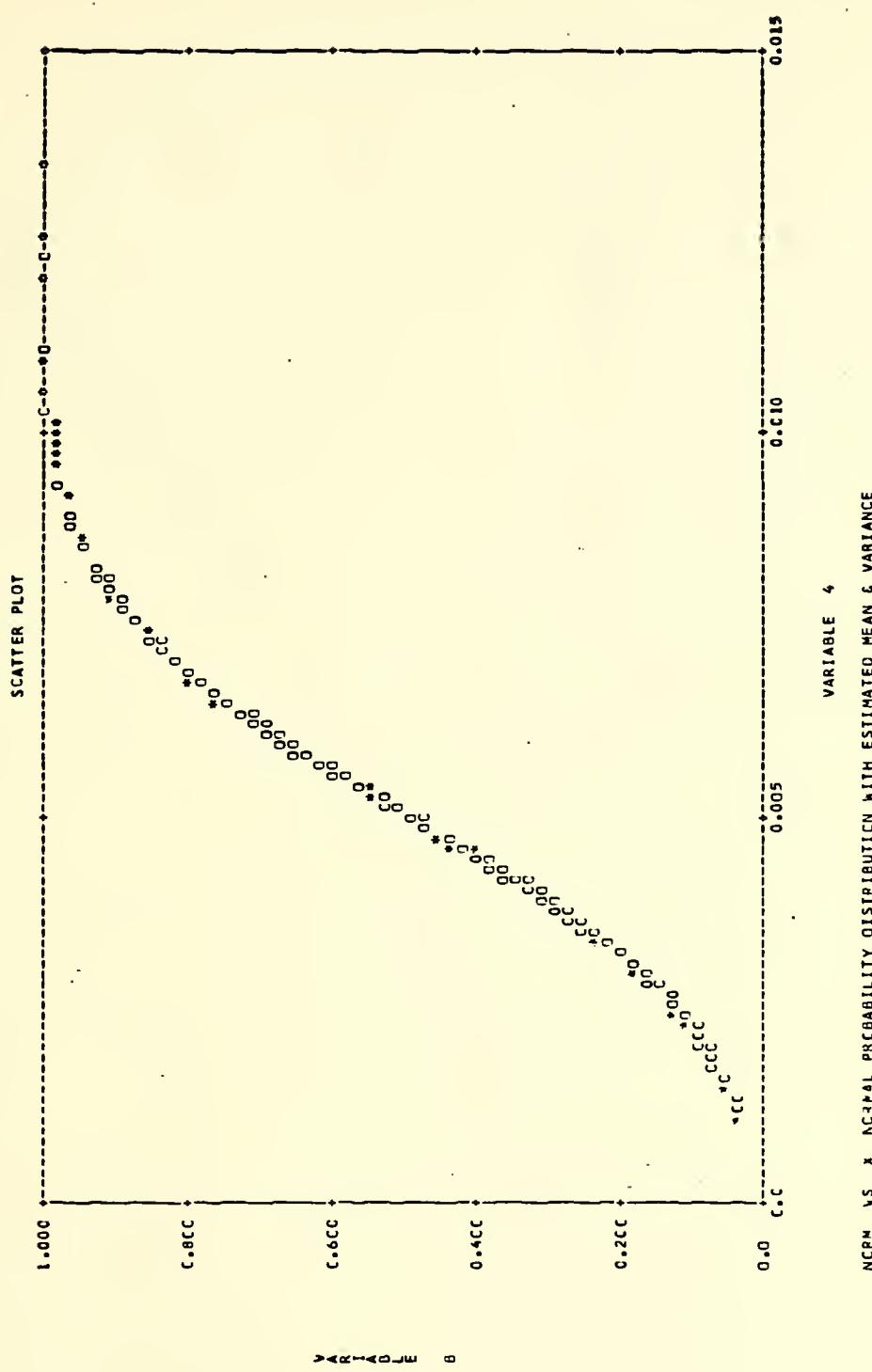




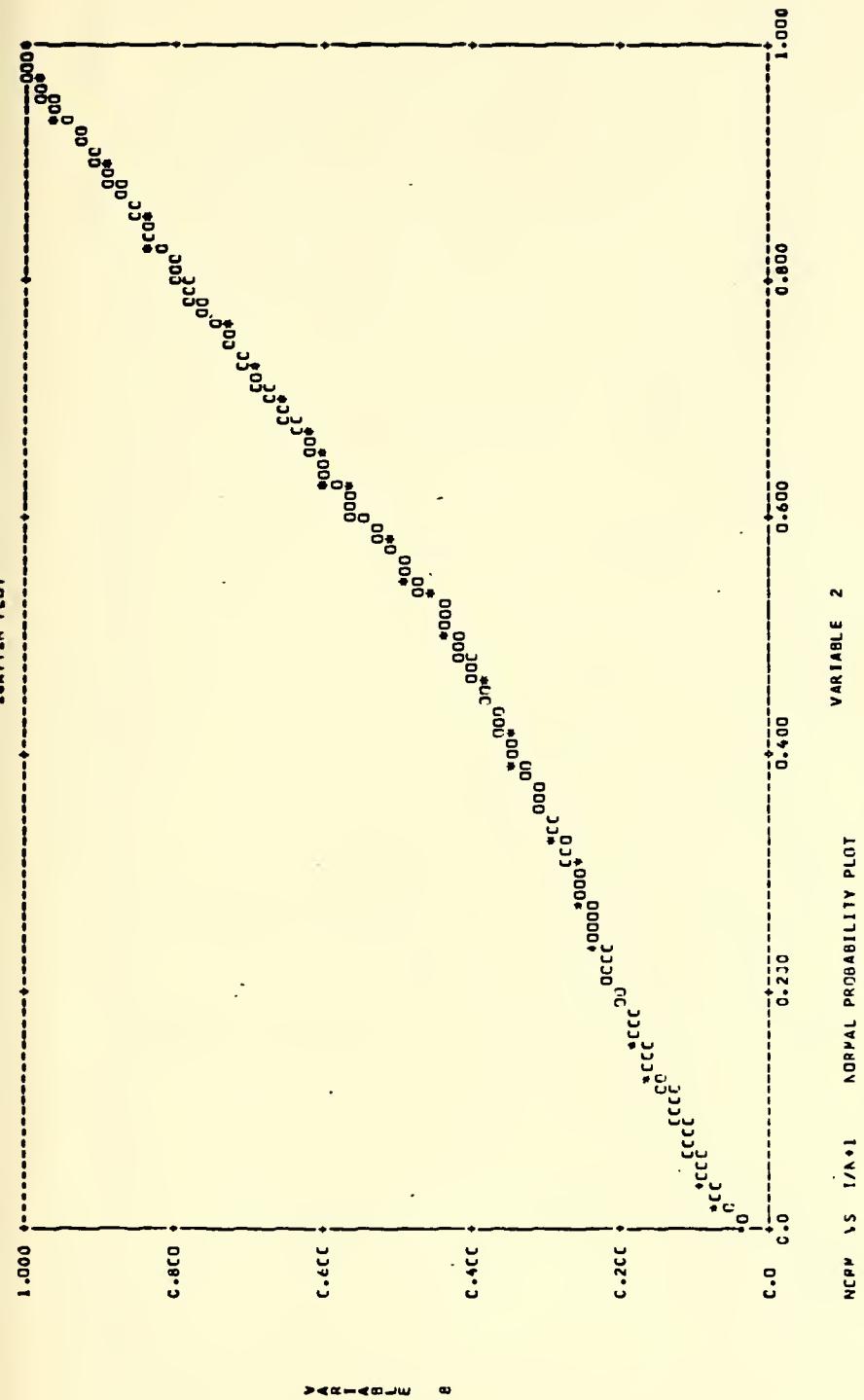


VARIABLE 2







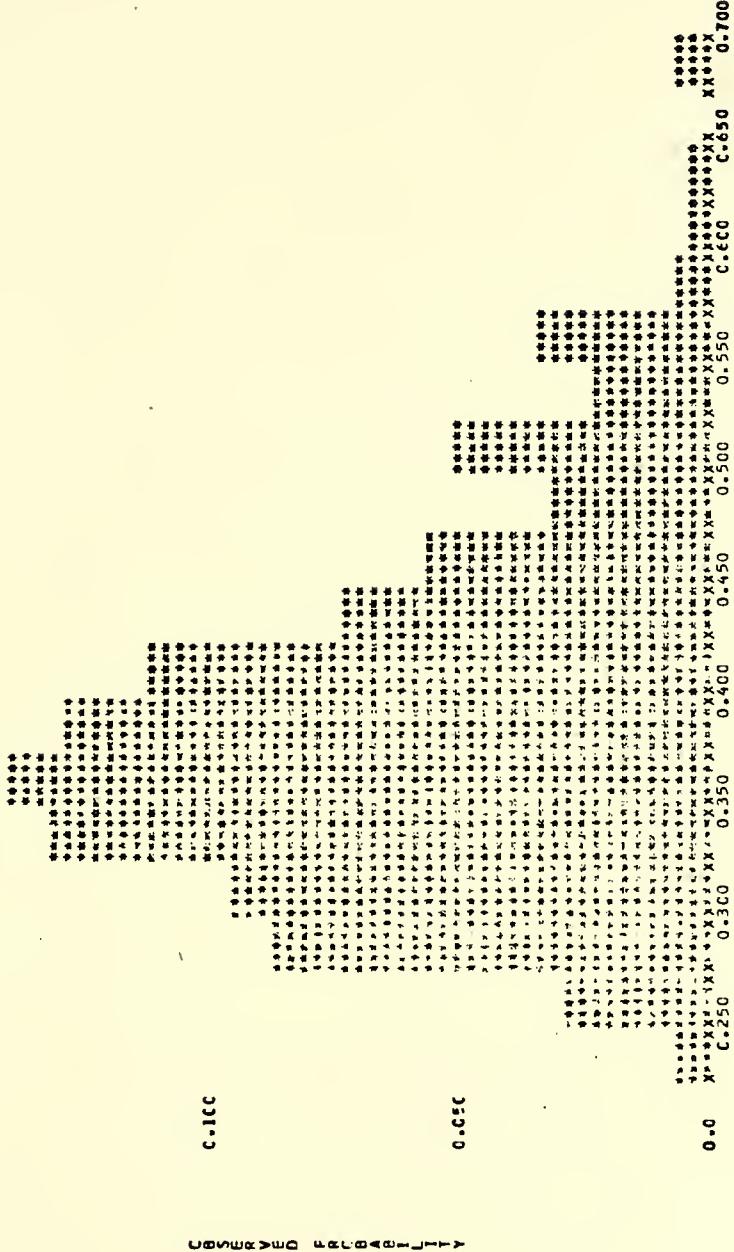






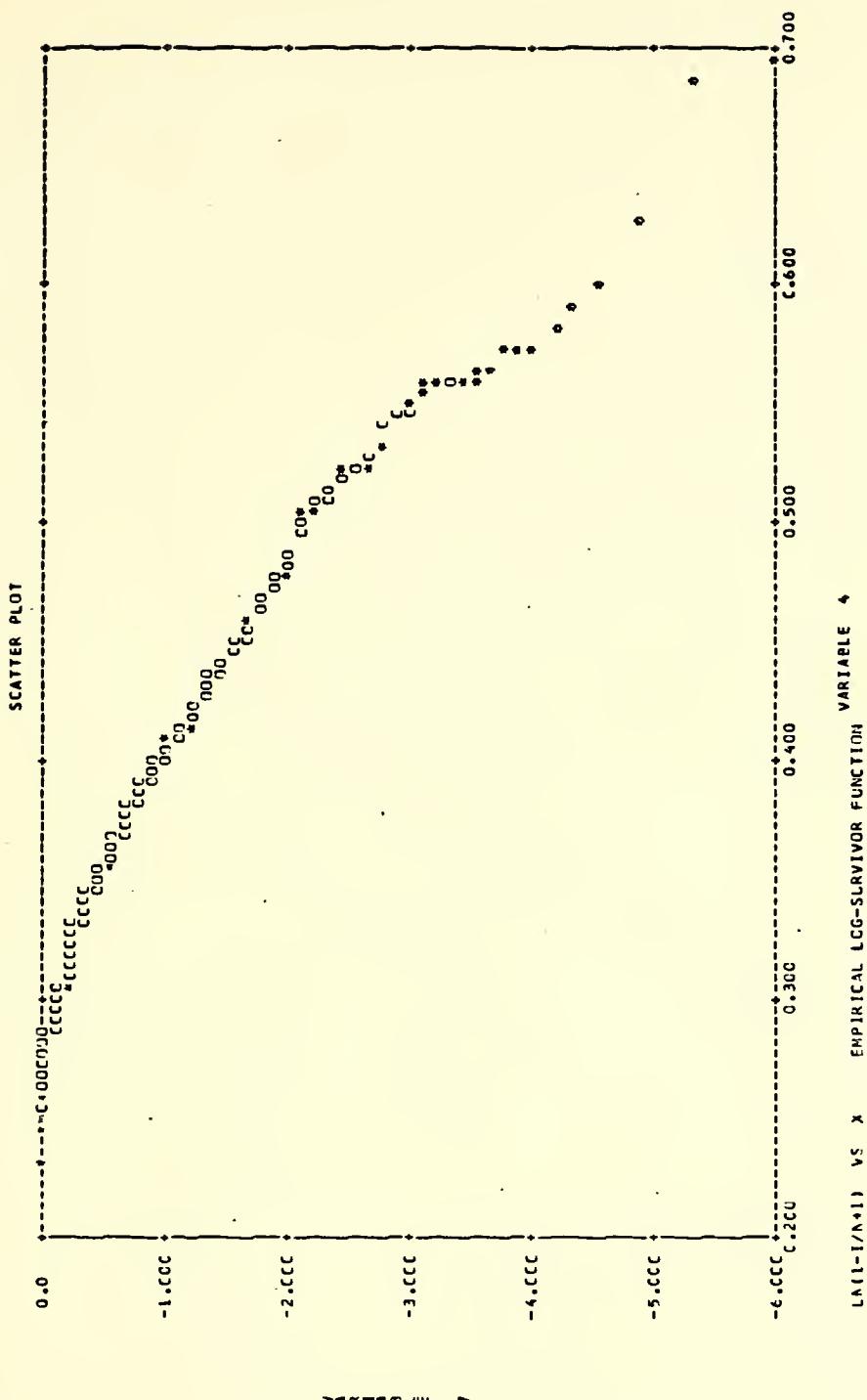


C.15C

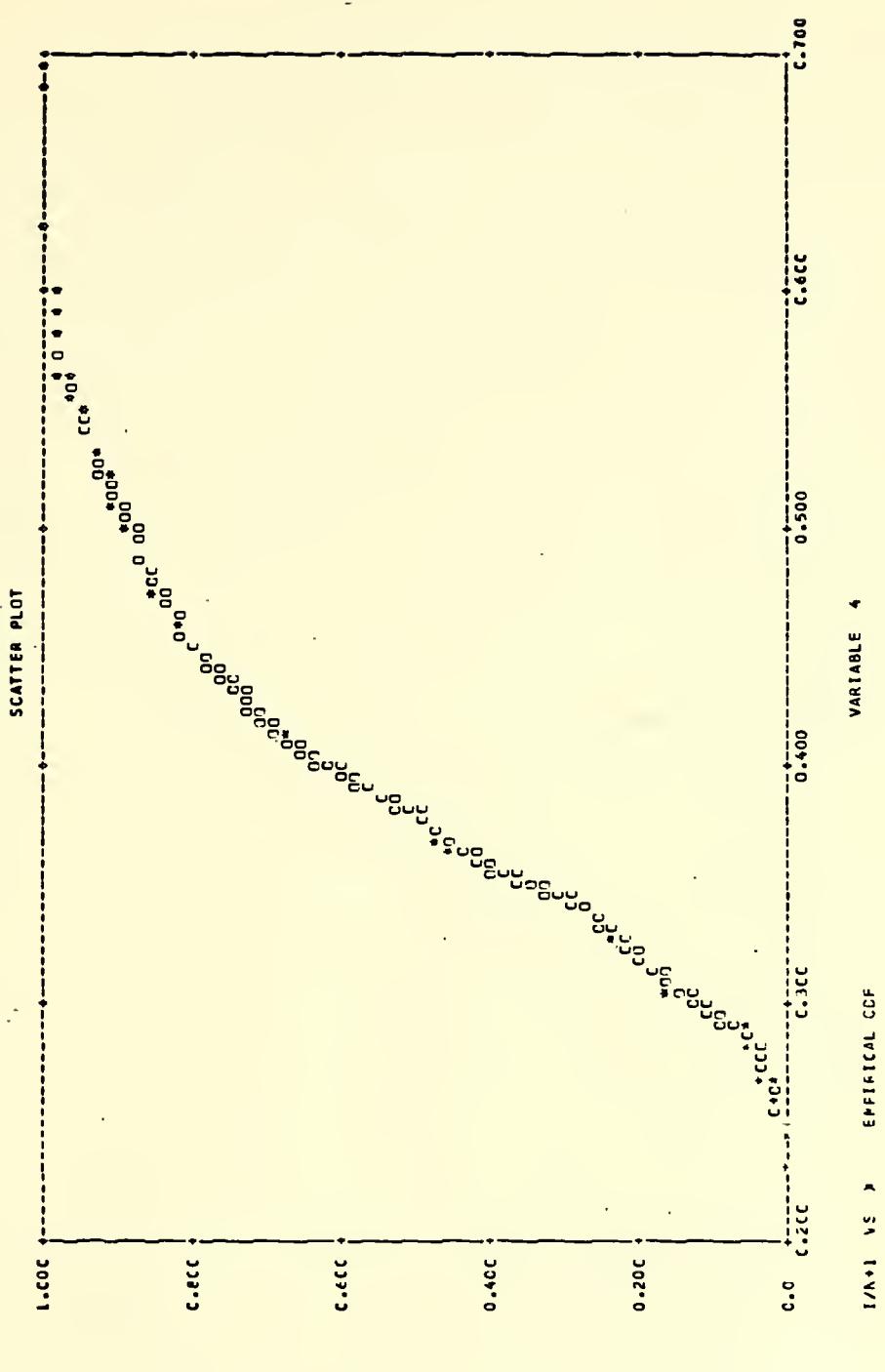


VARIABLE 4

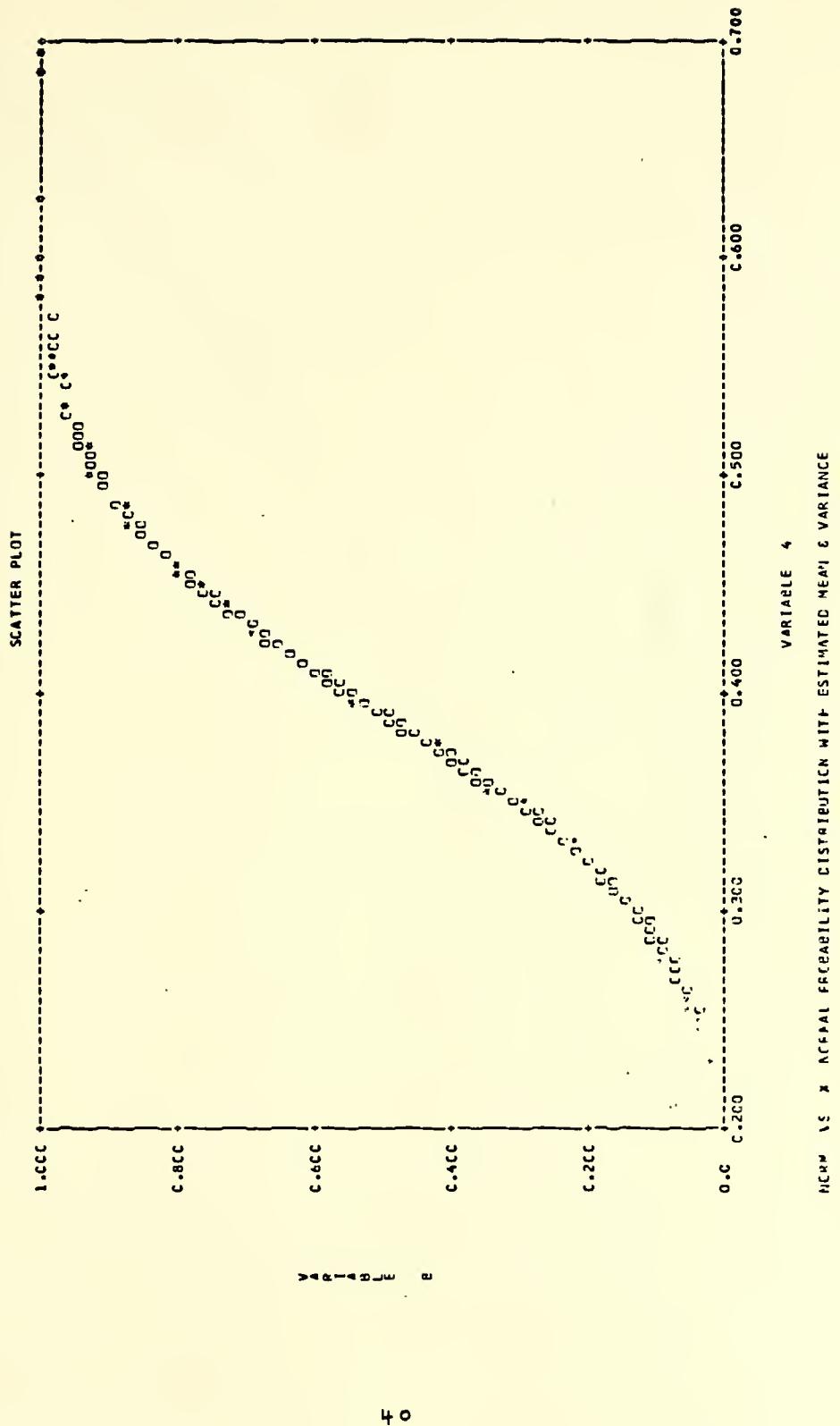






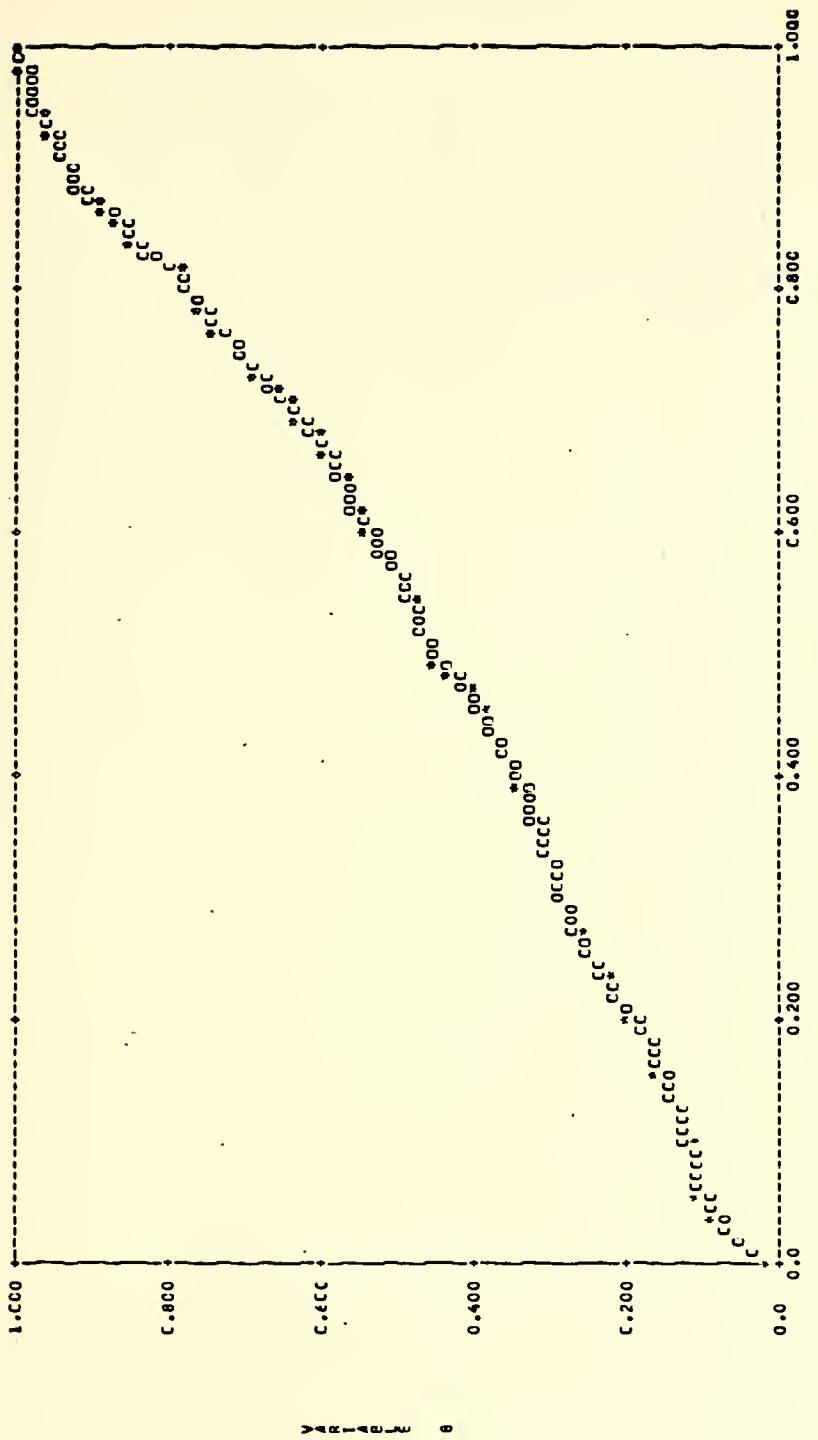








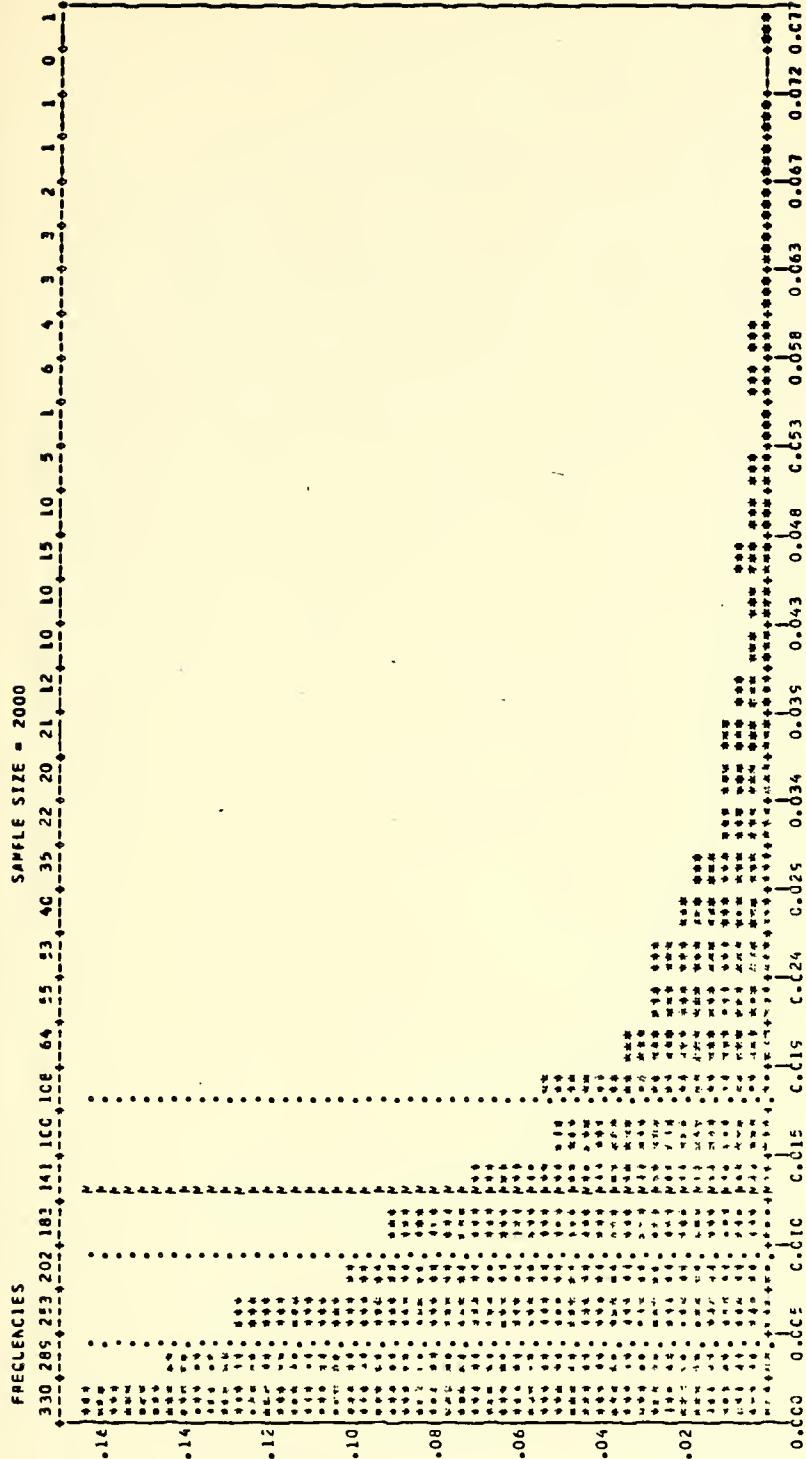
SCATTER PLOT



NCFM VS 1/N+1 NORMAL PROBABILITY PLOT

VARIABLE 2

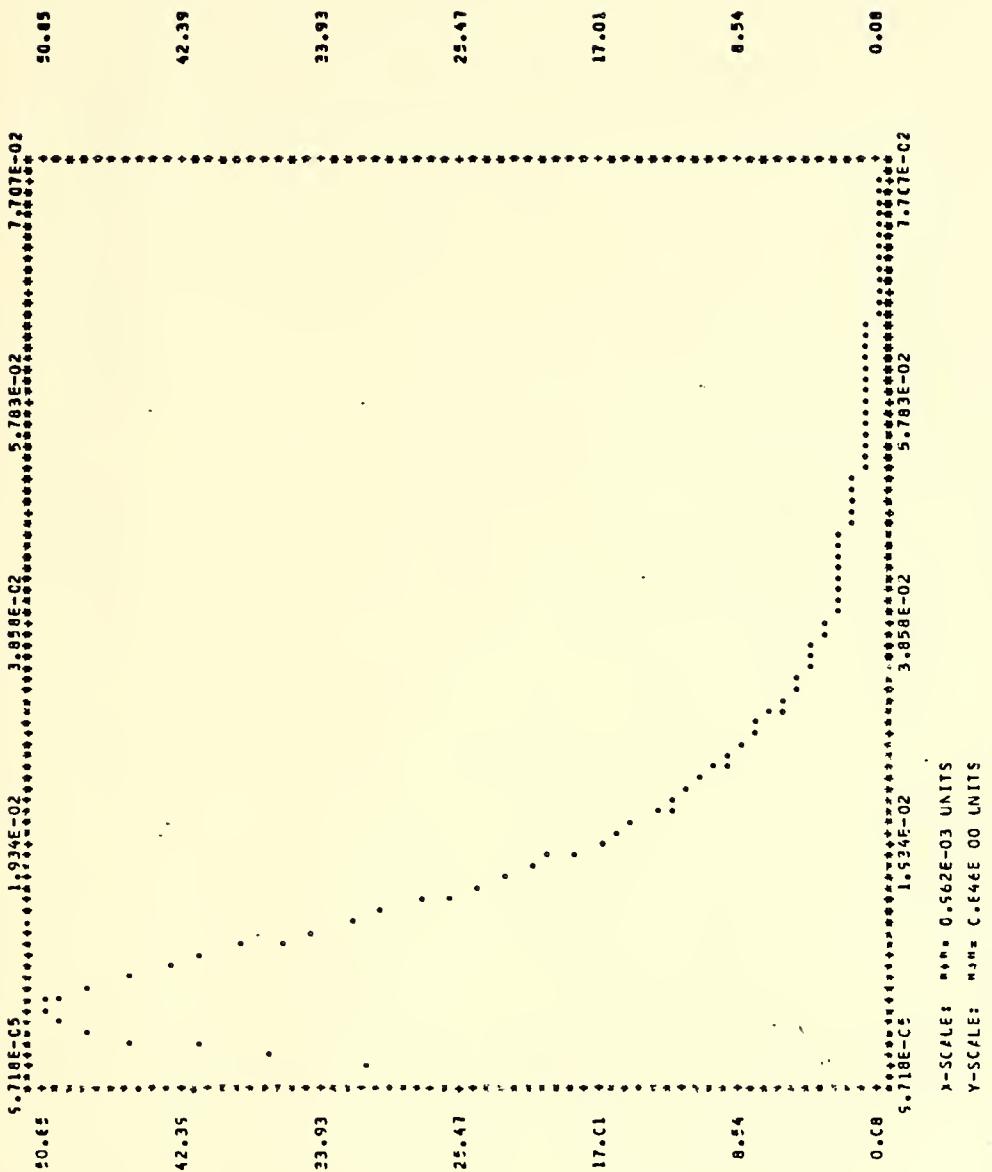




CENTRAL TENDENCY	HIGHER MOMENTS			DISTRIBUTION		
	VARIANCE	SKINNEDNESS	KURTOSIS	MINIMUM	10 QUANTILE	90 QUANTILE
MEAN	1.223714E-02	1.369225E-04	1.369225E-04	2.019138E-06	1.50644E-07	9.118356E-05
MEDIAN	8.552294E-03	1.167111E-02	9.50212E-01	1.661490E-00	1.02500E-00	1.15022E-03
MODE	9.68249E-03	9.50212E-01	7.653374E-02	3.606655E-00	2.50000E-00	3.68755E-03
RANGE	4.79155E-03	4.79155E-03	1.37787E-02	7.50000E-00	5.75000E-02	1.15568E-02
STDEV	7.64455E-03	7.64455E-03	2.58347E-02	1.90000E-00	1.25000E-00	2.75000E-02
COV	7.64455E-03	7.64455E-03	2.58347E-02	1.90000E-00	1.25000E-00	2.75000E-02
VAR	5.897915E-03	5.897915E-03	1.75000E-02	1.40000E-00	1.00000E-00	1.70000E-02
SKEWNESS	1.50000E-03	1.50000E-03	4.00000E-02	3.00000E-00	2.00000E-00	3.50000E-02
KURTOSIS	3.606655E-03	3.606655E-03	1.00000E-02	7.50000E-00	5.00000E-00	1.00000E-02
SPREAD	3.583454E-03	3.583454E-03	9.00000E-02	6.00000E-00	4.00000E-00	7.00000E-02
SPREADAC	3.583454E-03	3.583454E-03	9.00000E-02	6.00000E-00	4.00000E-00	7.00000E-02

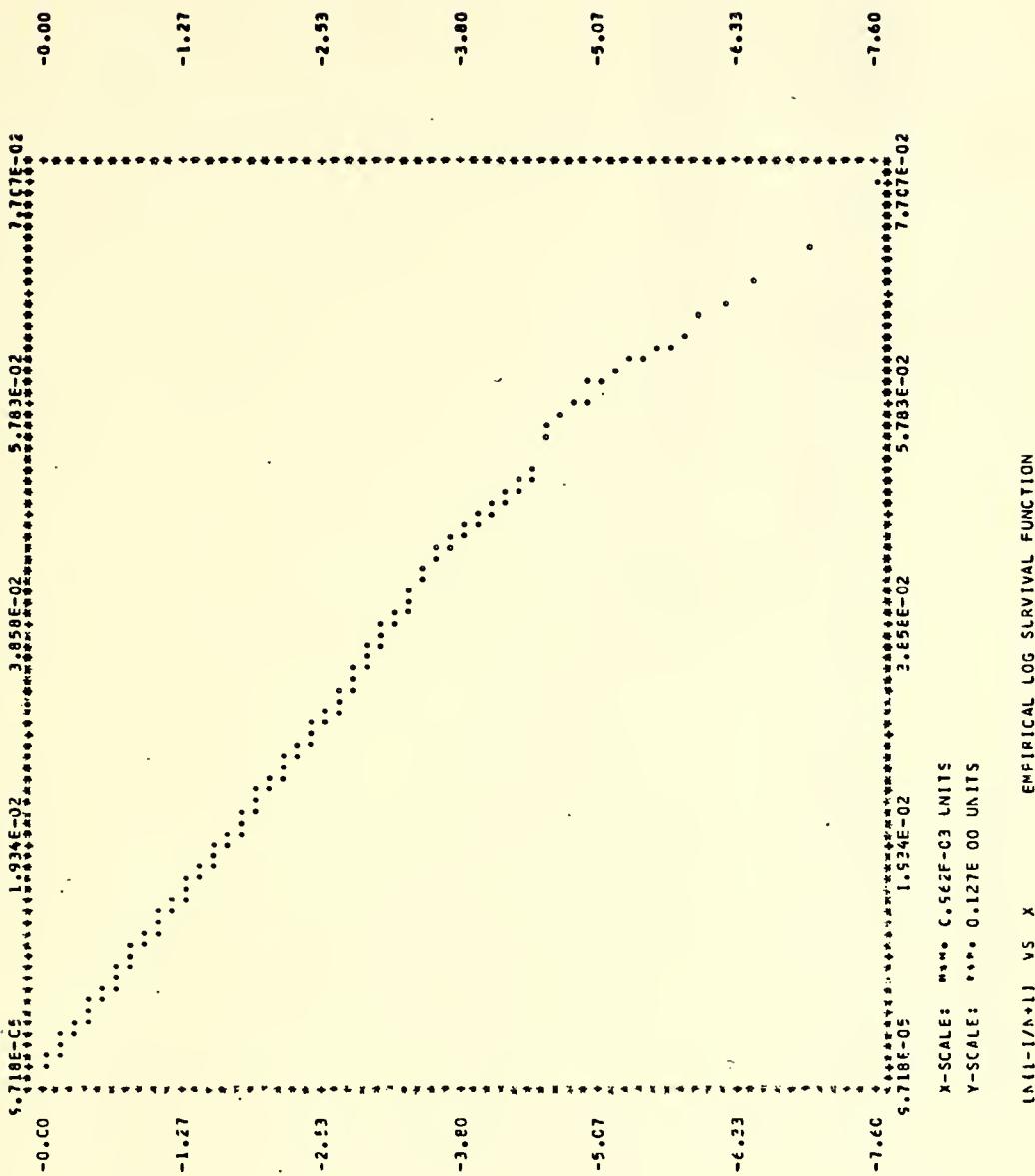
INTERAL SCUFFE NORM SAMPLE SIZE  $N = 2CCC$   
UNIFORM RANGE VARIABLE SAMPLE SIZE  $N = 100$   
TRIANGULAR WINDOW BANDWIDTH =  $3/\sqrt{NIN}$



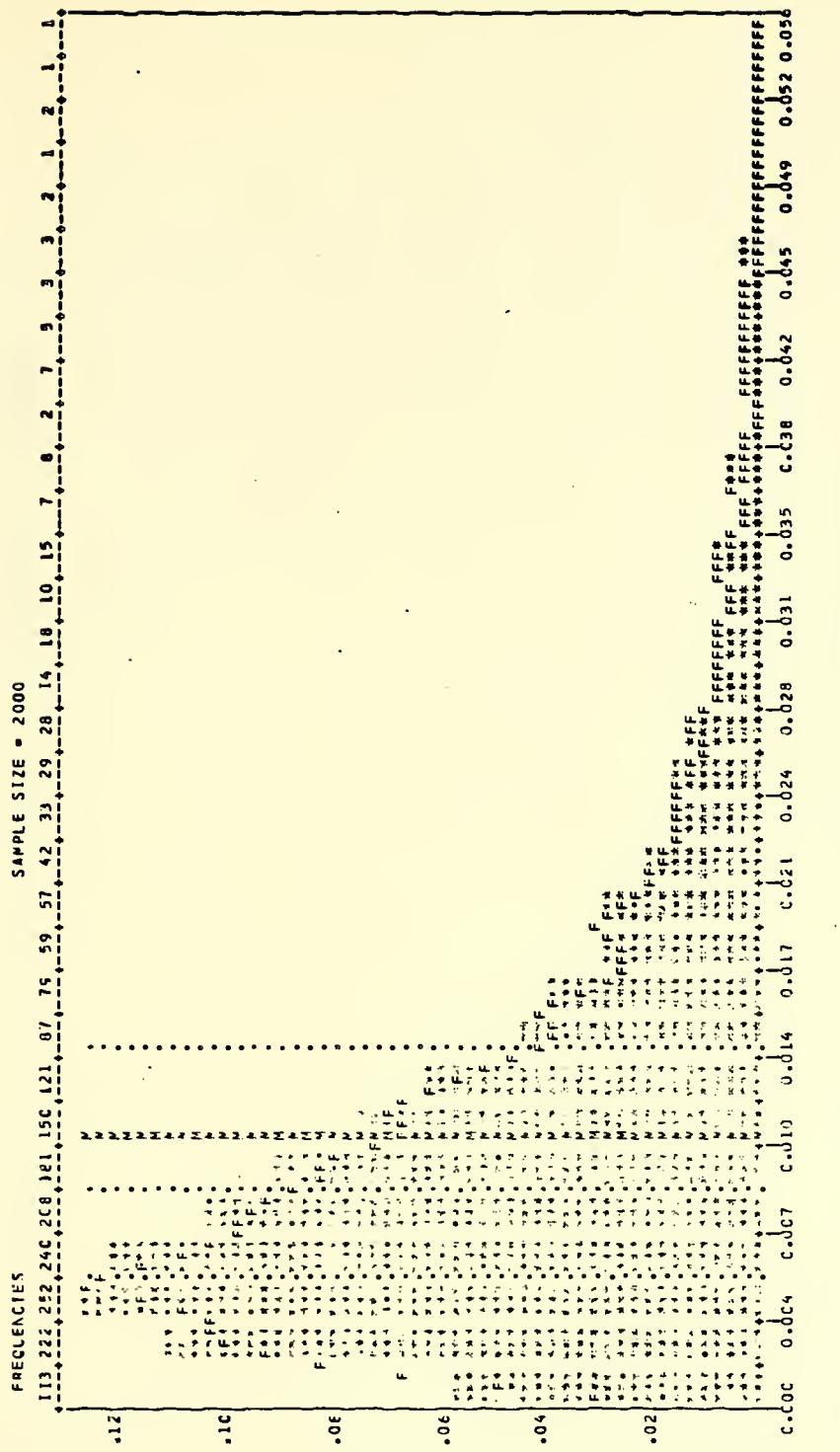


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
INTEGRAL SCALE OF NCFP SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
ESTIMATED CENSITY FUNCTION IS EVALUATED AT 1000 EQUALLY SPACED POINTS









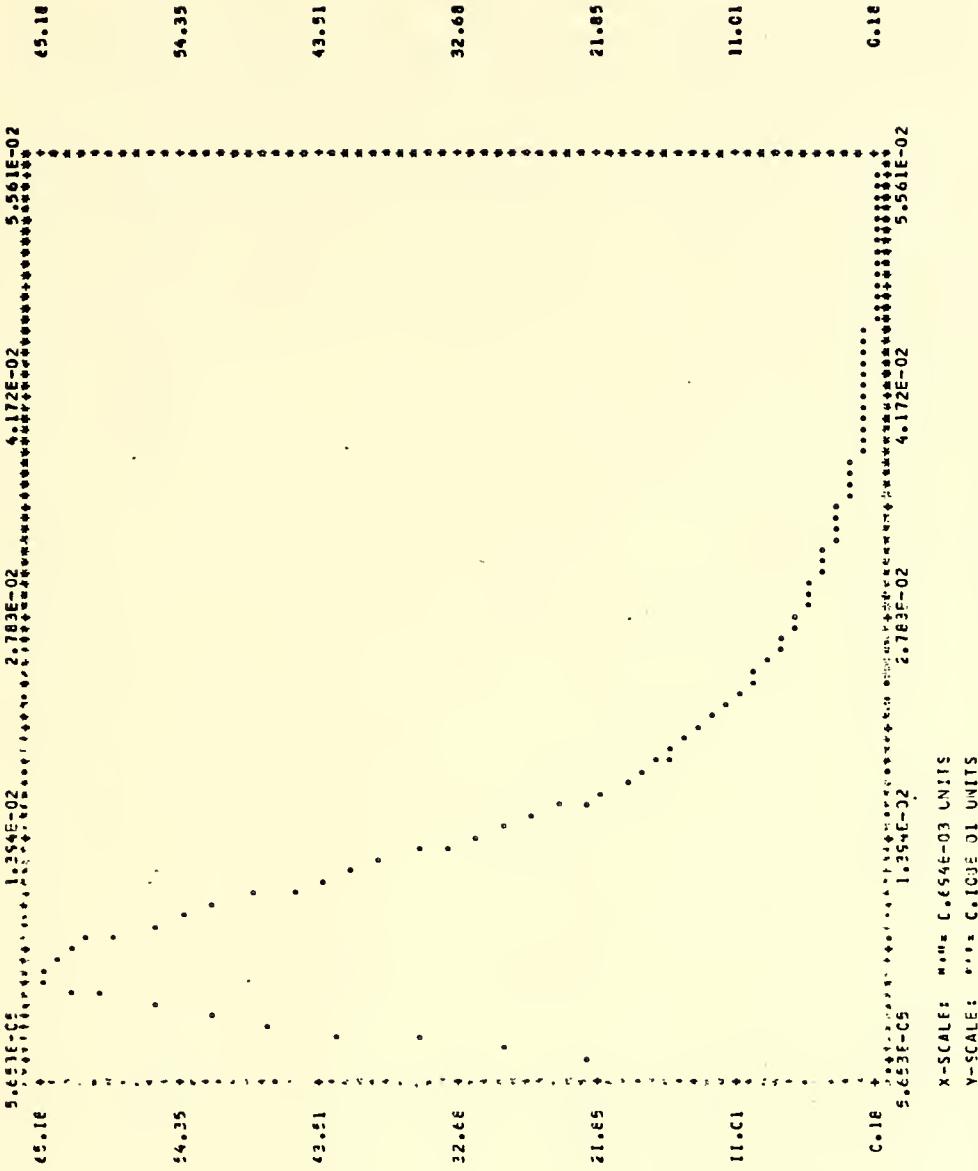
45

CENTRAL TENDENCY	S P R E D E C	H I G H E R C E N T R A L M O M E N T S	D I S T R I B U T I O N
MEAN	7.24816E-02	M3 9.666197E-07	MINIMUM 2.652262E-05
MEAN	8.37455E-02	M4 1.255779E-08	0.10 QUANTILE 2.532429E-03
TRIMMED MEAN	8.35954E-02	SKEWNESS 1.617755E-00	0.25 QUANTILE 4.752459E-03
TRIMMED MEAN	8.37221E-02	KURTOSIS 3.209741E-00	0.50 QUANTILE 8.474532E-03
GEOGRAPHIC MEAN	7.66125E-02	M11SPEAC 1.75 QUANTILE 1.442532E-02	0.75 QUANTILE 2.224433E-02
HARM. MEAN	4.64555E-02	M12 0.9049 0.049 0.052 0.056	MAXIMUM 2.56C843E-02

INTEGRAL SCALING WITH SAMPLE SIZE  $n = 2000$   
 UNIFORM RANDOM VARIABLE SAMPLE SIZE  $N = 200$   
 TRIANGULAR KERNEL.

BRANCH CUT = 3/SQRT(N)

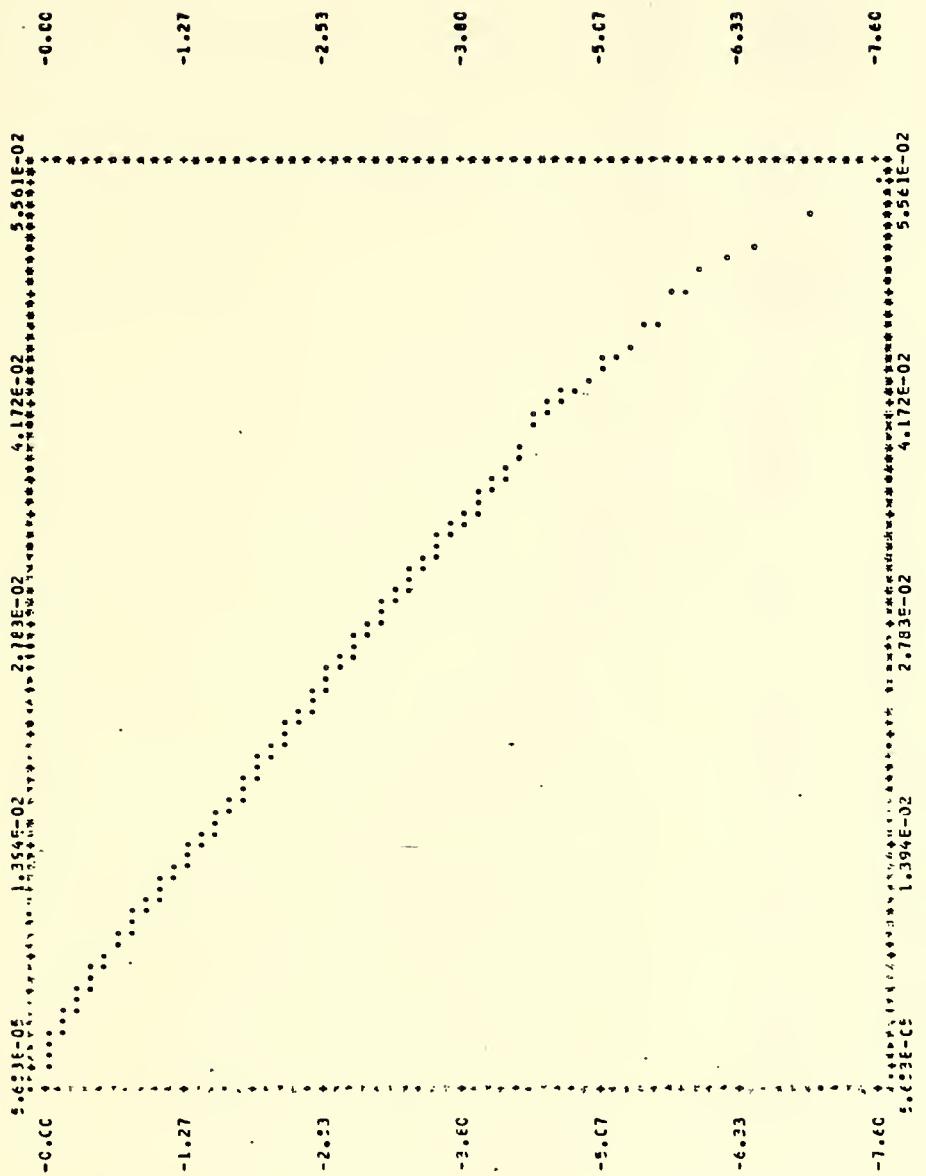




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ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NECRN RANGE VARIABLE  
 INTEGRAL SCALE FROM SAMPLE SIZE = 2000      TRIANGULAR WINDOW      BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 1000 EQUALLY SPACED POINTS

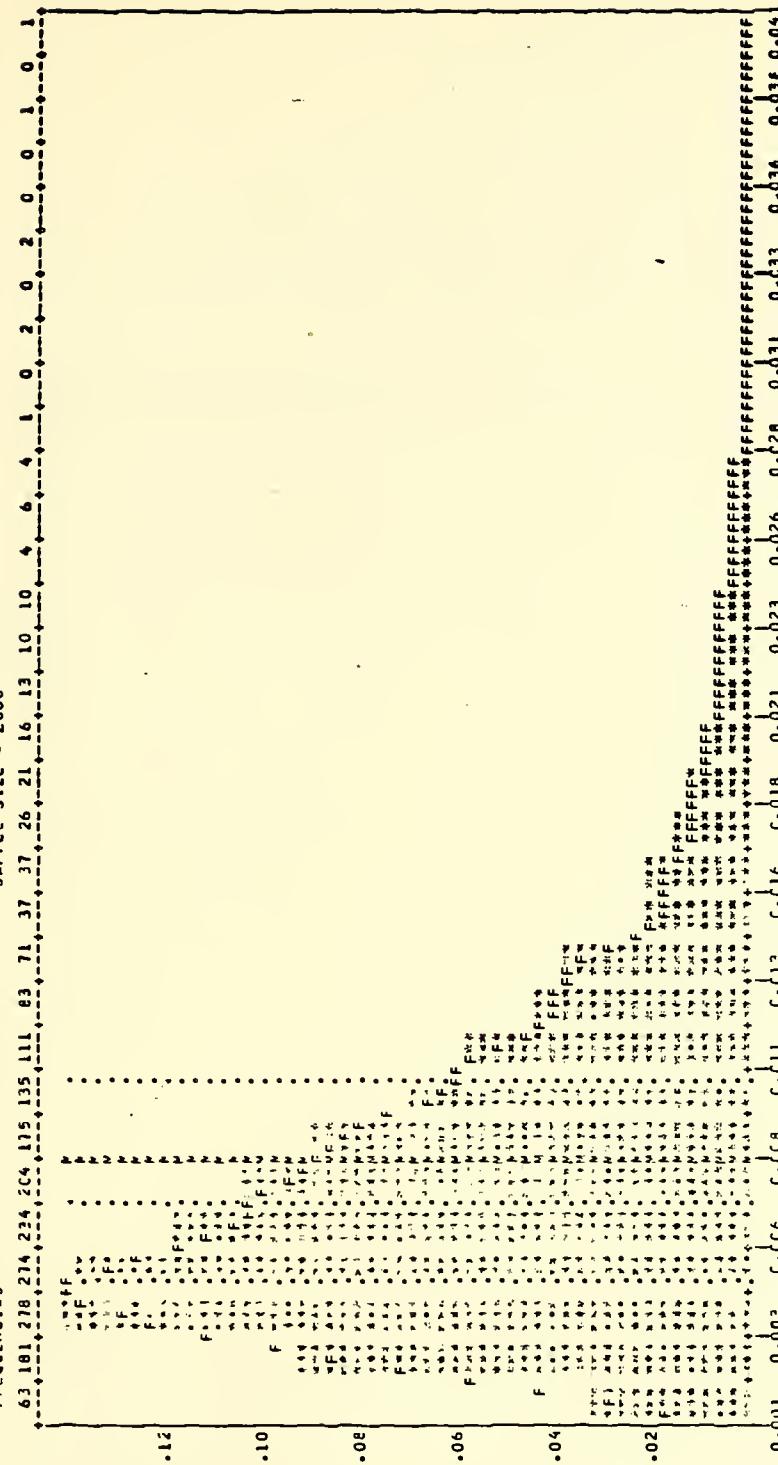






## FREQUENCIES

SAMPLE SIZE = 2000



48

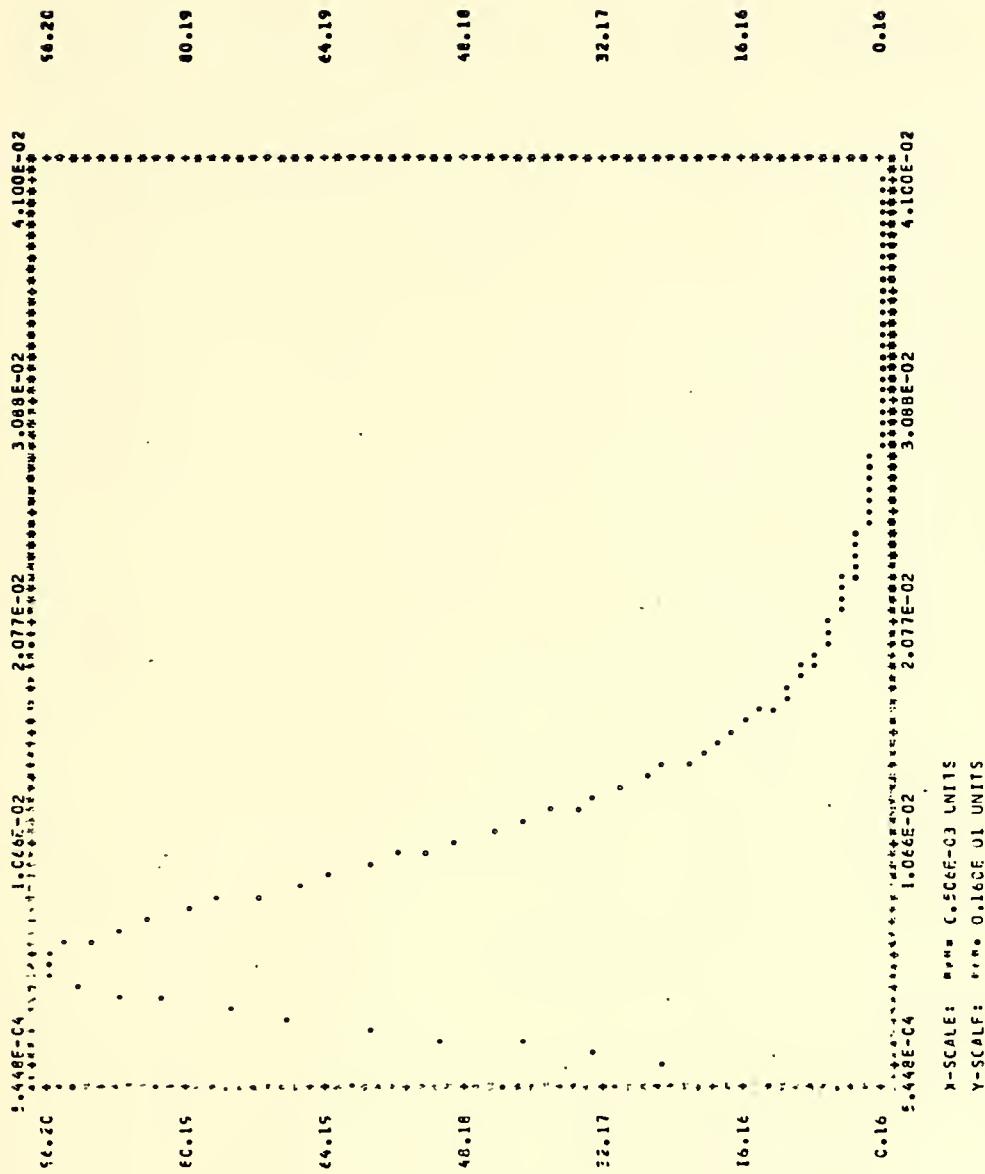
## CENTRAL TENDENCY

	SPREAD	HIGHER CENTRAL MOMENTS	DISTRIBUTION
MEAN	7.071304E-03	2.55686E-05	M3 2.013204E-07
MEDIAN	6.676704E-03	5.06644E-03	M4 4.377604E-09
TRIMMED MEAN	6.557244E-03	6.42155E-01	SKENNESS 1.557135E-03
WICHMAN	6.442247E-03	4.04516E-02	KURTOSIS 3.6966078E-00
GECM	6.412266E-03	6.013735E-03	(MECHANICAL)
HARM. MEAN	5.184553E-03		90 QUANTILE 1.03964E-02
			MAXIMUM 1.429808E-02
			4.05948E-02

	SPREAD	HIGHER CENTRAL MOMENTS	DISTRIBUTION
MINIMUM	0.013204E-07	M3 2.013204E-07	4.46914E-04
QUANTILE	0.377604E-09	M4 4.377604E-09	2.82603E-03
75 QUANTILE	1.557135E-03	SKENNESS 1.557135E-03	4.24600E-03
90 QUANTILE	3.6966078E-00	KURTOSIS 3.6966078E-00	1.03964E-02
		(MECHANICAL)	1.429808E-02
		90 QUANTILE 1.03964E-02	4.05948E-02

INTEGRAL SCALPE NORM SAMPLE SIZE M = 2000  
UNIFORM RANGE VARIABLE SAMPLE SIZE N = 500  
TRIANGULAR WINDOW.      ERANCHICHT = 3/SQRT(N)



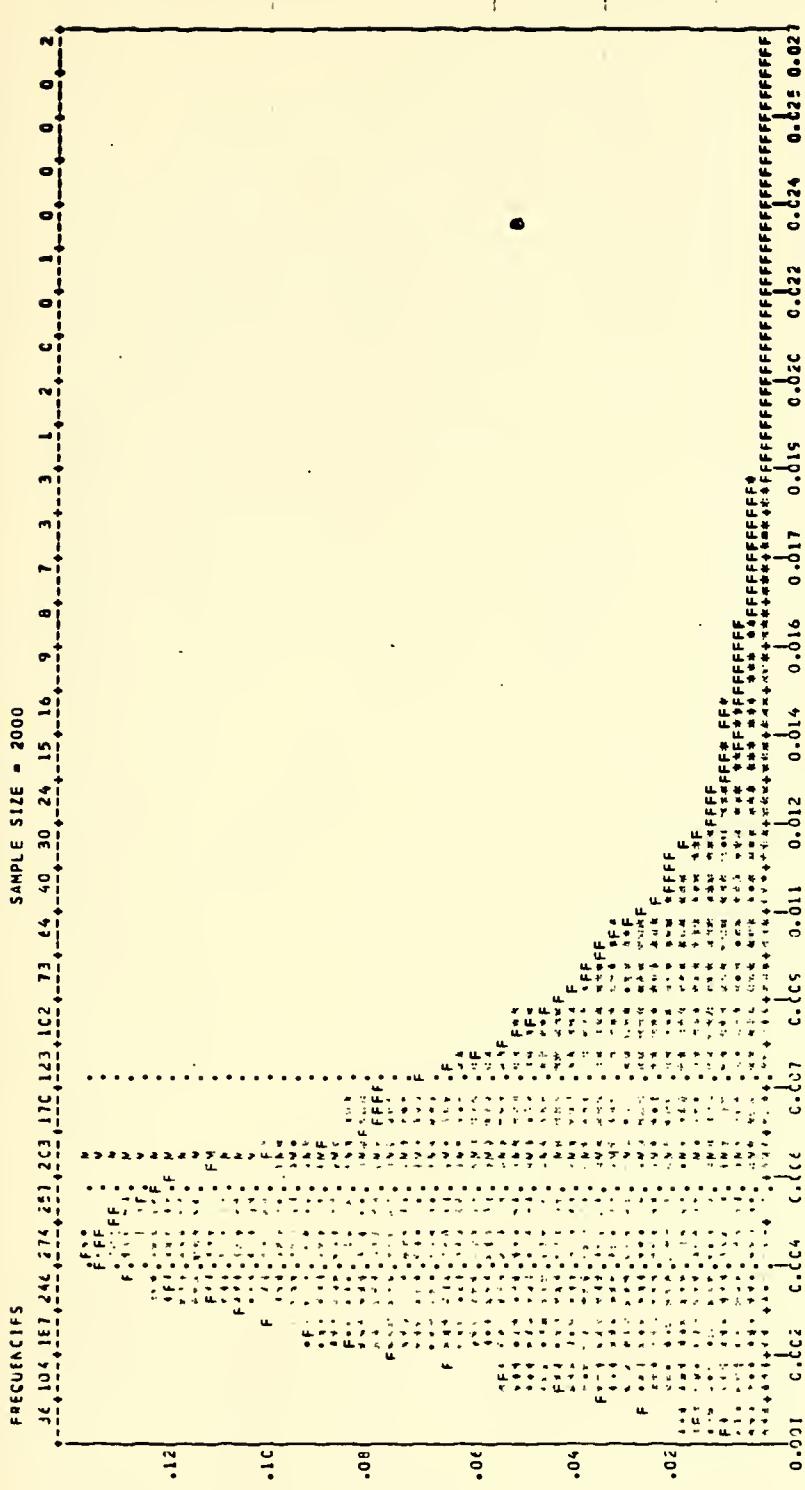


ESTIMATED CENSITY FUNCTION VS INTEGRAL SQUARE ACROSS RANDOM VARIABLE  
 INTEGRAL SQUARE ACROSS SAMPLE SIZE = 2000      TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED CENSITY FLNCTIN IS EVALUATED AT 100 EQUALLY SPACED POINTS





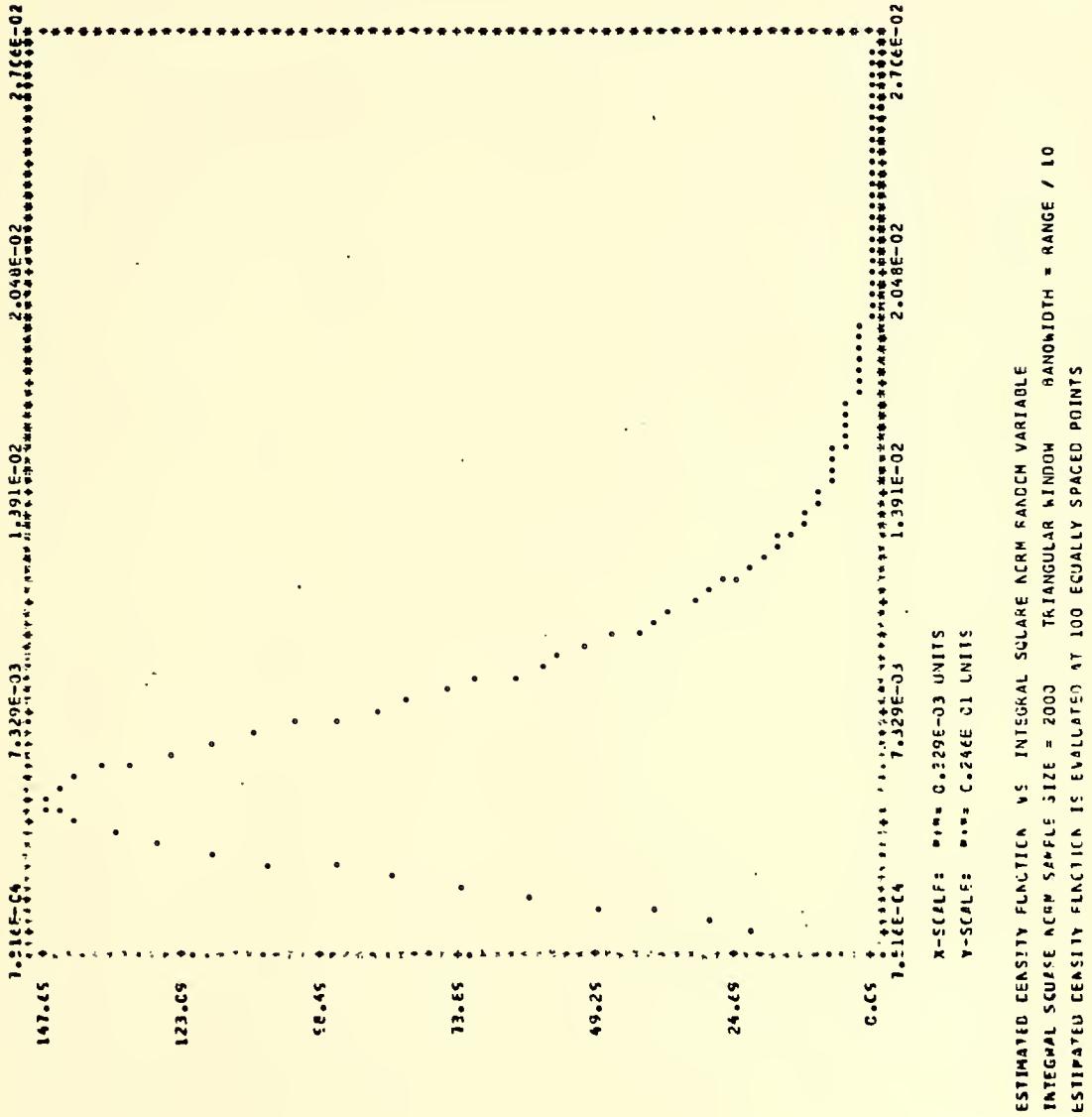




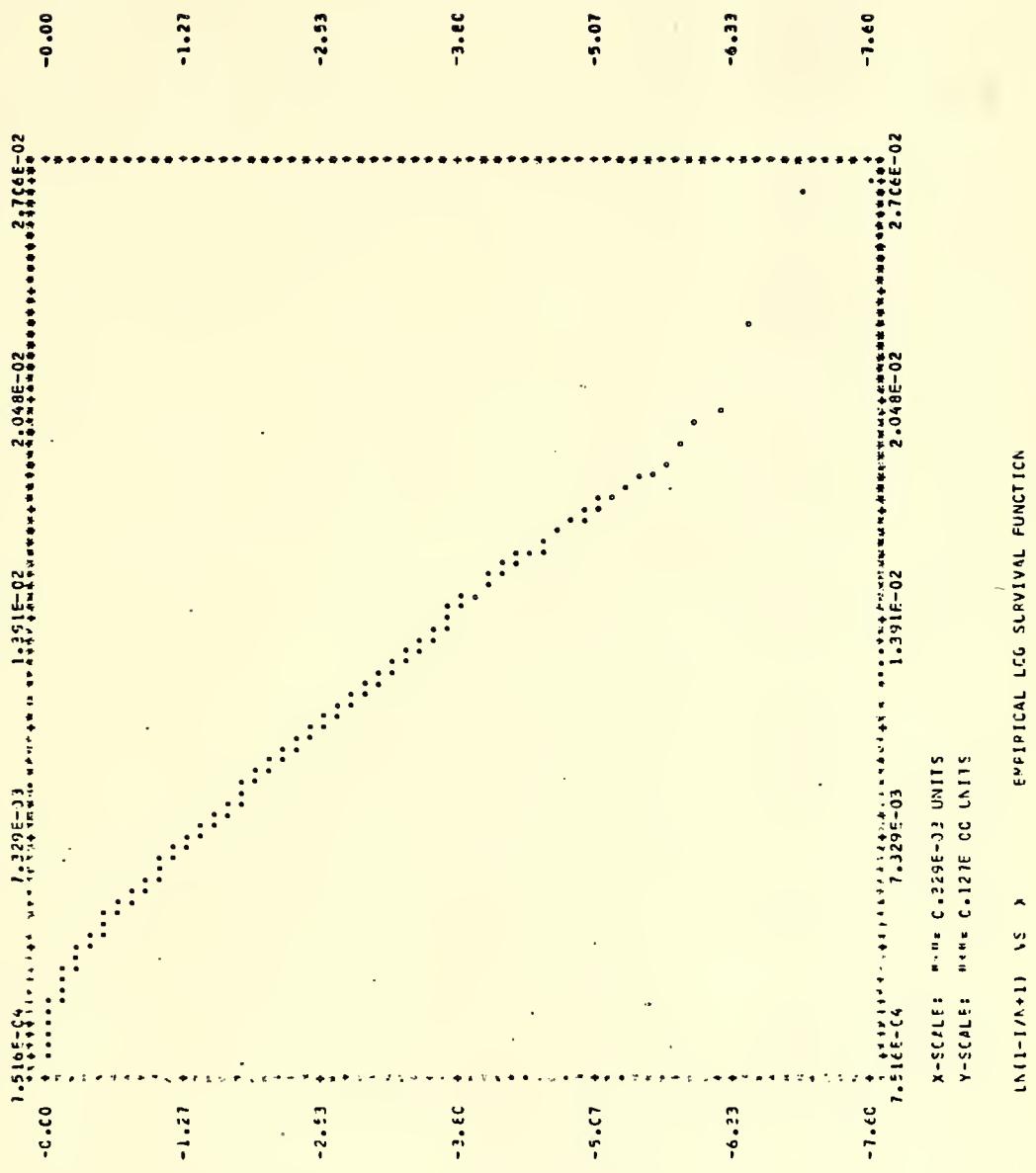
CENTRAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS				DISTRIBUTION
		VARIANCE	M <sup>3</sup>	M <sup>4</sup>	KURTOSIS	
MEAN	S <sup>2</sup> S <sup>2</sup> C <sup>2</sup> E <sup>-1</sup> -C <sup>2</sup>	1.823811E-05	4.418073E-08	6.358899E-00	1.438565E-00	UNIFORM
MEDIAN	E <sup>-1</sup> S <sup>2</sup> C <sup>2</sup> E <sup>-1</sup> -C <sup>2</sup>	1.823811E-05	4.418073E-08	6.358899E-00	1.438565E-00	UNIFORM
TRIMMED MEAN	E <sup>-1</sup> S <sup>2</sup> C <sup>2</sup> E <sup>-1</sup> -C <sup>2</sup>	1.823811E-05	4.418073E-08	6.358899E-00	1.438565E-00	UNIFORM
MEAN	S <sup>2</sup> S <sup>2</sup> C <sup>2</sup> E <sup>-1</sup> -C <sup>2</sup>	1.823811E-05	4.418073E-08	6.358899E-00	1.438565E-00	UNIFORM
STDEV	E <sup>-1</sup> S <sup>2</sup> C <sup>2</sup> E <sup>-1</sup> -C <sup>2</sup>	1.823811E-05	4.418073E-08	6.358899E-00	1.438565E-00	UNIFORM
STDEV <sup>2</sup>	E <sup>-1</sup> S <sup>2</sup> C <sup>2</sup> E <sup>-1</sup> -C <sup>2</sup>	1.823811E-05	4.418073E-08	6.358899E-00	1.438565E-00	UNIFORM
CHIEF VAR	E <sup>-1</sup> S <sup>2</sup> C <sup>2</sup> E <sup>-1</sup> -C <sup>2</sup>	1.823811E-05	4.418073E-08	6.358899E-00	1.438565E-00	UNIFORM
RANGE	E <sup>-1</sup> S <sup>2</sup> C <sup>2</sup> E <sup>-1</sup> -C <sup>2</sup>	2.630198E-02	3.64113E-03	3.58958CE-00	1.438565E-00	UNIFORM
WIDESPREAD	E <sup>-1</sup> S <sup>2</sup> C <sup>2</sup> E <sup>-1</sup> -C <sup>2</sup>	2.630198E-02	3.64113E-03	3.58958CE-00	1.438565E-00	UNIFORM
GEOM MEAN	E <sup>-1</sup> S <sup>2</sup> C <sup>2</sup> E <sup>-1</sup> -C <sup>2</sup>	1.823811E-05	4.418073E-08	6.358899E-00	1.438565E-00	UNIFORM
FARM MEAN	4.576621E-C2	1.823811E-05	4.418073E-08	6.358899E-00	1.438565E-00	UNIFORM

INTEGRAL SQUARE NORM SAMPLE SIZE  $N = 2000$   
 UNIFORM RANGE VARIABLE SAMPLE SIZE  $N \approx 1000$   
 TRIANGULAR BRANCH SUBJECTS  $\approx 3/5(2000)$





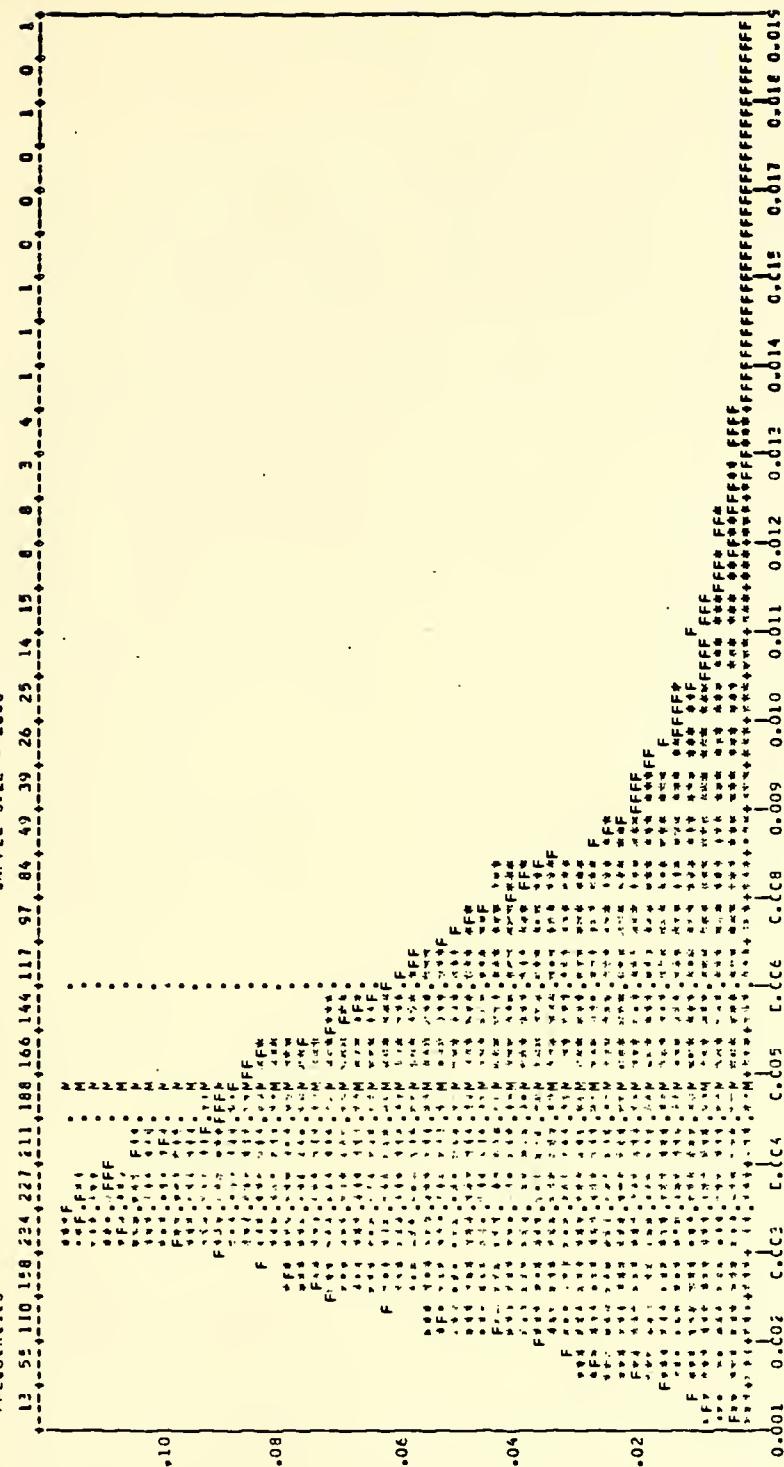






## FREQUENCIES

SAMPLE SIZE = 2000



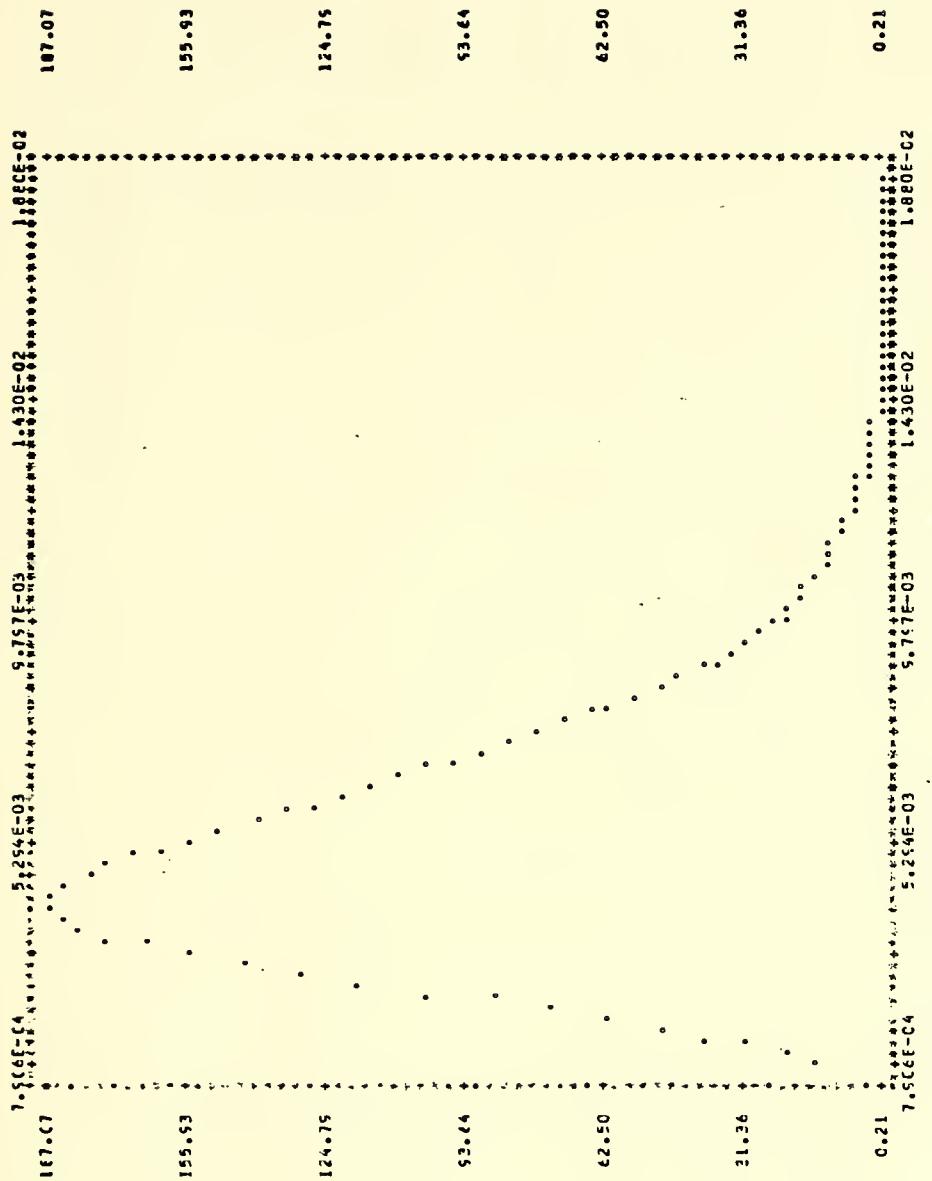
## CENTRAL TENDENCY

	SPLAO	HIGHER CENTRAL MOMENTS	DISTRIBUTION
MEAN	5.110052E-02	M3 5.192715E-02	MINIMUM 1.0221252E-08
MEDIAN	4.155115E-02	M4 2.718815E-02	10 QUANTILE 1.026120E-10
TRIMMED MEAN	4.156155E-02	COEF VAR 4.455291E-02	25 QUANTILE 1.03037E-00
MICRO MEAN	4.174122E-02	FACTOR 1.004267E-02	50 QUANTILE 1.030240E-00
CECH MEAN	4.172422E-02	MOMENT 2.924502E-03	MEAN 4.000000E-03
FAPP MEAN	4.162222E-02	SKURTOSIS 1.680240E-03	10 QUANTILE 4.000000E-03

INTEGRAL SQUARE NEW SAMPLE SIZE  $n = 2000$   
UNIFCP (ANCCP VARIABLE SAMPLE SIZE  $n = 1500$ )  
TRIANGULAR WINOCH. FANCWICP = 3/SCFT (n)

	SPLAO	HIGHER CENTRAL MOMENTS	DISTRIBUTION
MEAN	5.110052E-02	M3 5.192715E-02	MINIMUM 1.0221252E-08
MEDIAN	4.155115E-02	M4 2.718815E-02	10 QUANTILE 1.026120E-10
TRIMMED MEAN	4.156155E-02	COEF VAR 4.455291E-02	25 QUANTILE 1.03037E-00
MICRO MEAN	4.174122E-02	FACTOR 1.004267E-02	50 QUANTILE 1.030240E-00
CECH MEAN	4.172422E-02	MOMENT 2.924502E-03	MEAN 4.000000E-03
FAPP MEAN	4.162222E-02	SKURTOSIS 1.680240E-03	10 QUANTILE 4.000000E-03

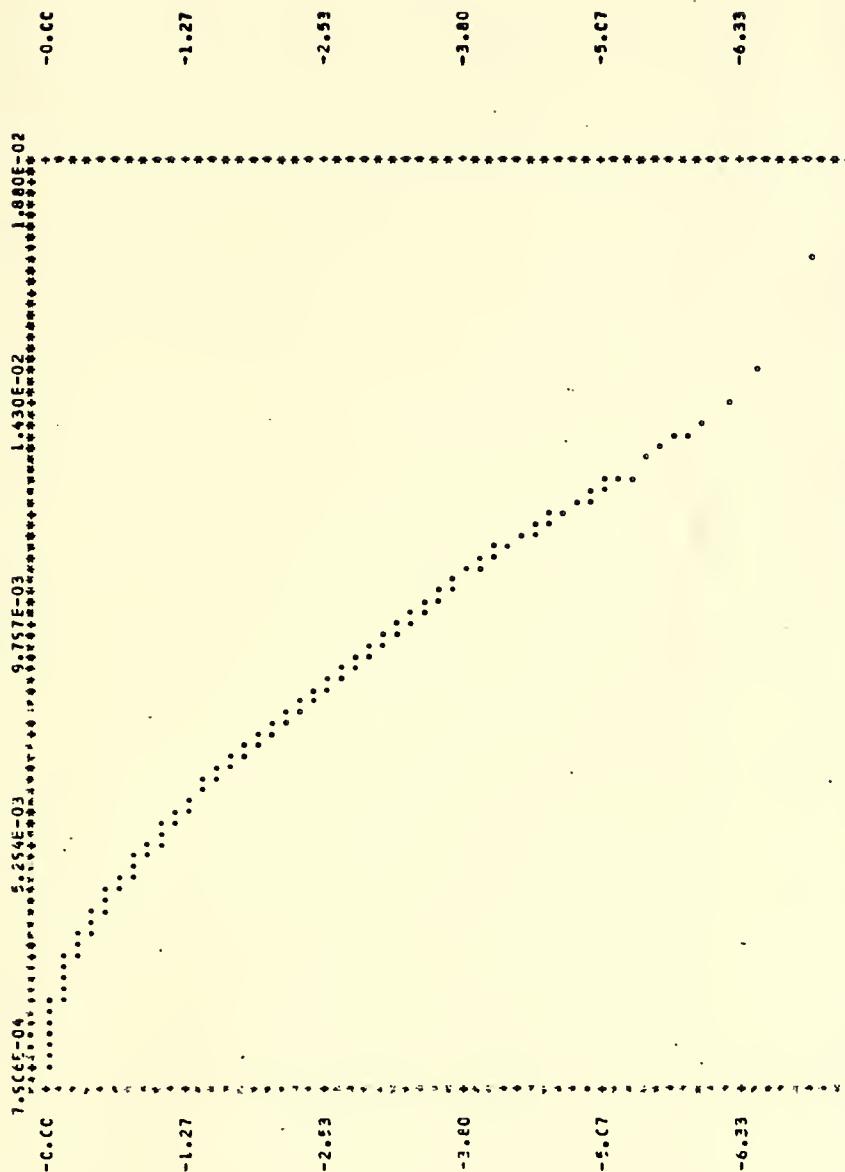




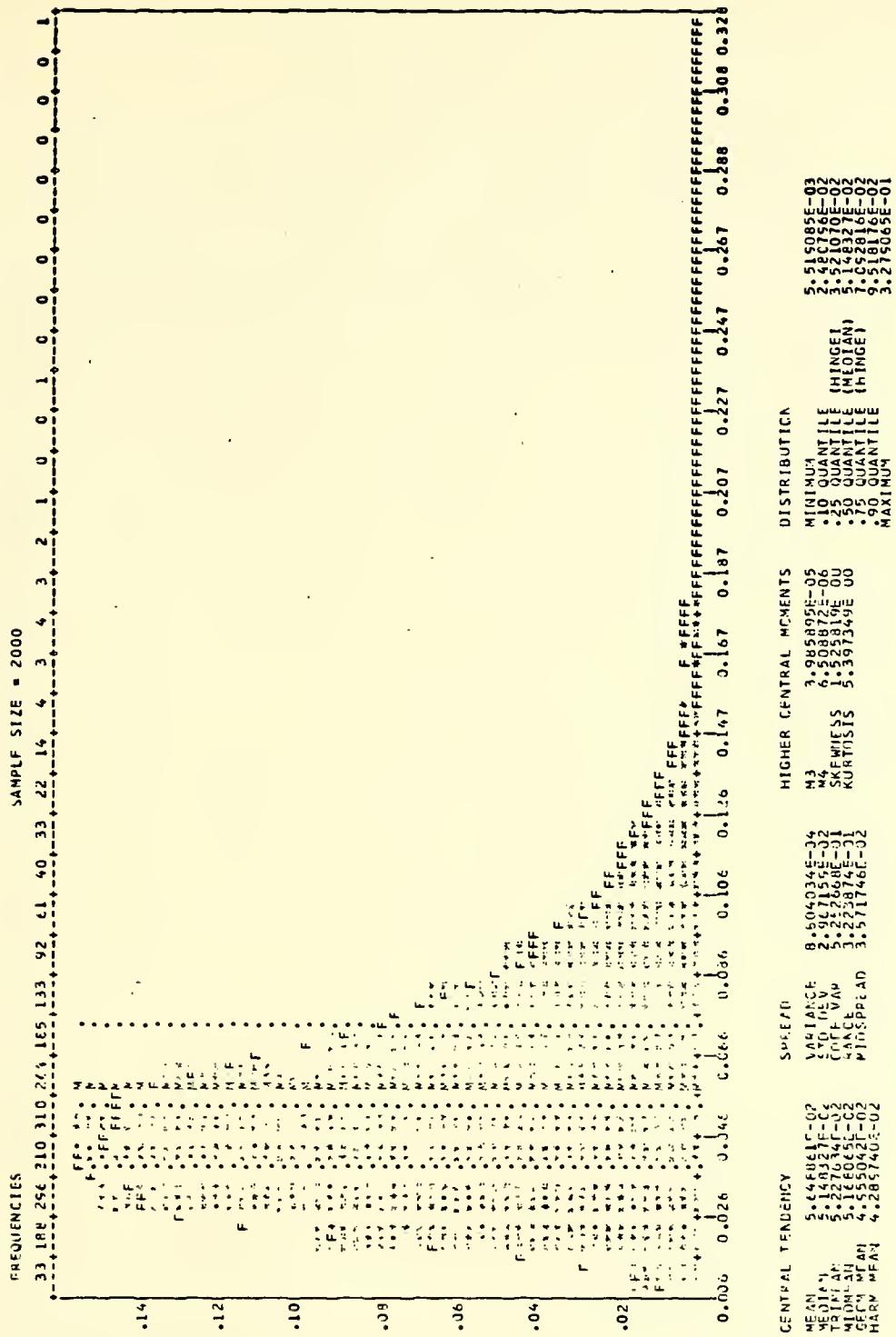
ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000    TRIANGULAR WINDOW    BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 1000 EQUALLY SPACED POINTS



LN(1/N(t)) VS X                                    EMPIRICAL LCG SURVIVAL FUNCTION

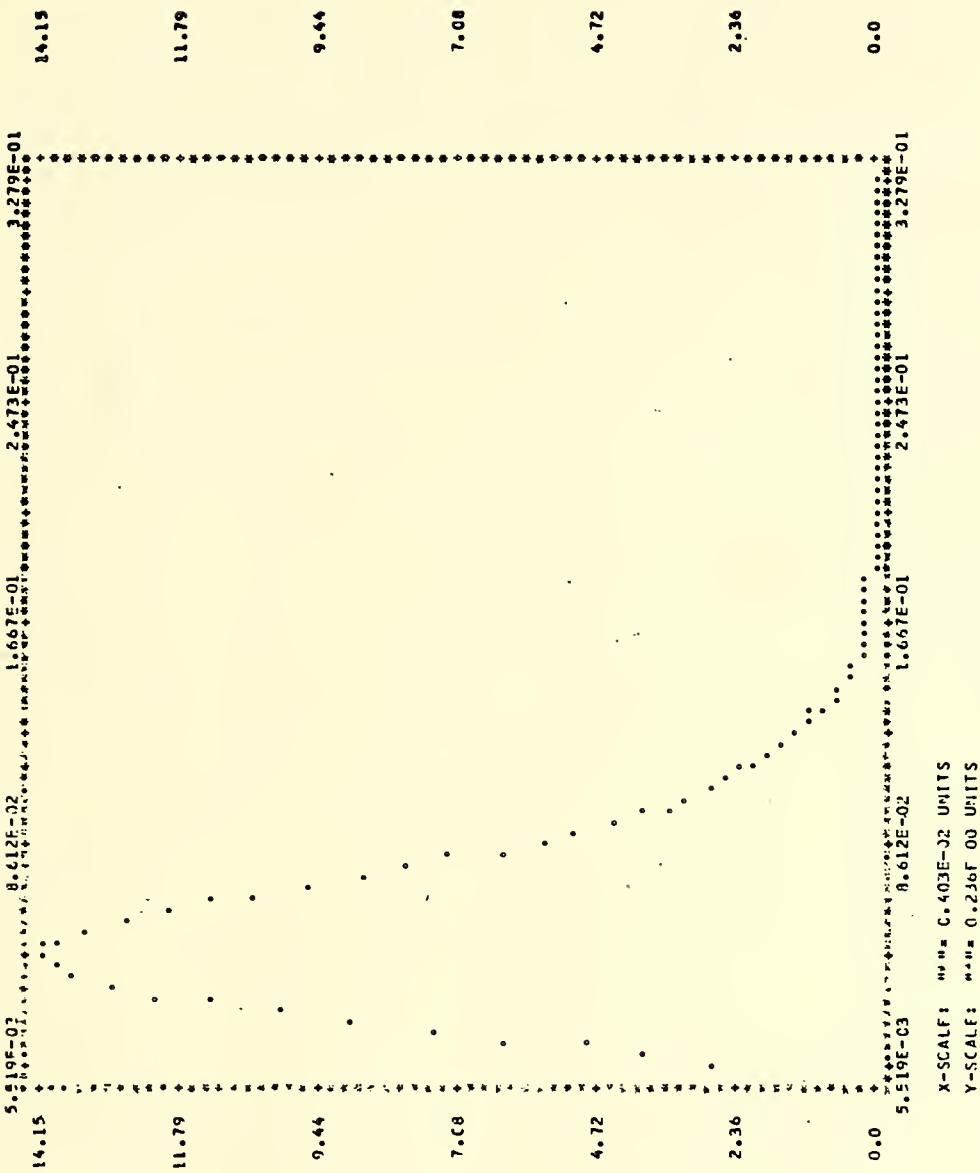






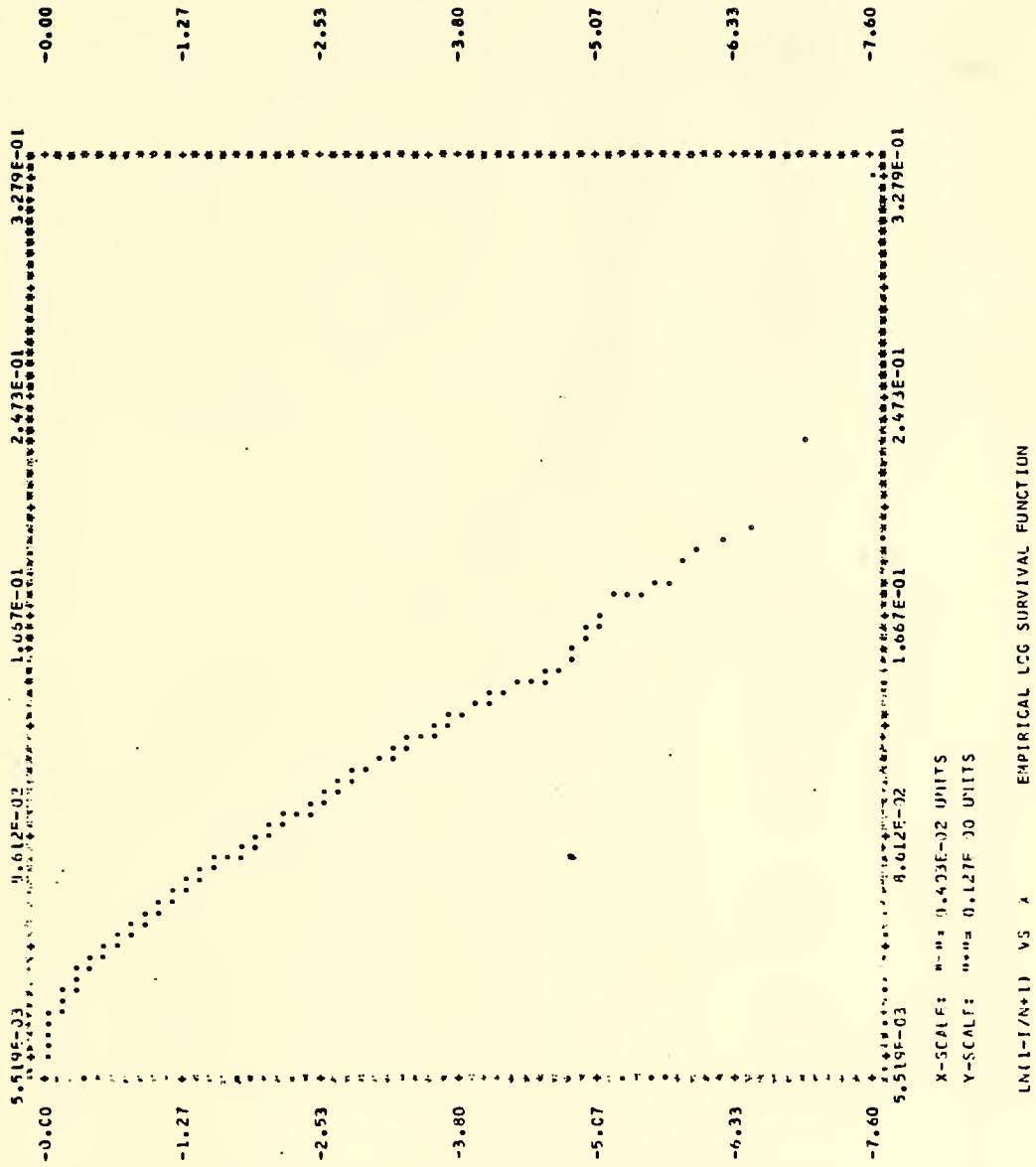
INTEGRAL SQUARES WITH SAMPLE SIZE  $n = 2000$   
 UNIFORM RANDOM VARIABLE SAMPLE SIZE  $n = 100$   
 TRIANGULAR WINDOW.  
 $\text{ENVELOPE} = 1/\text{SQT}(n)$



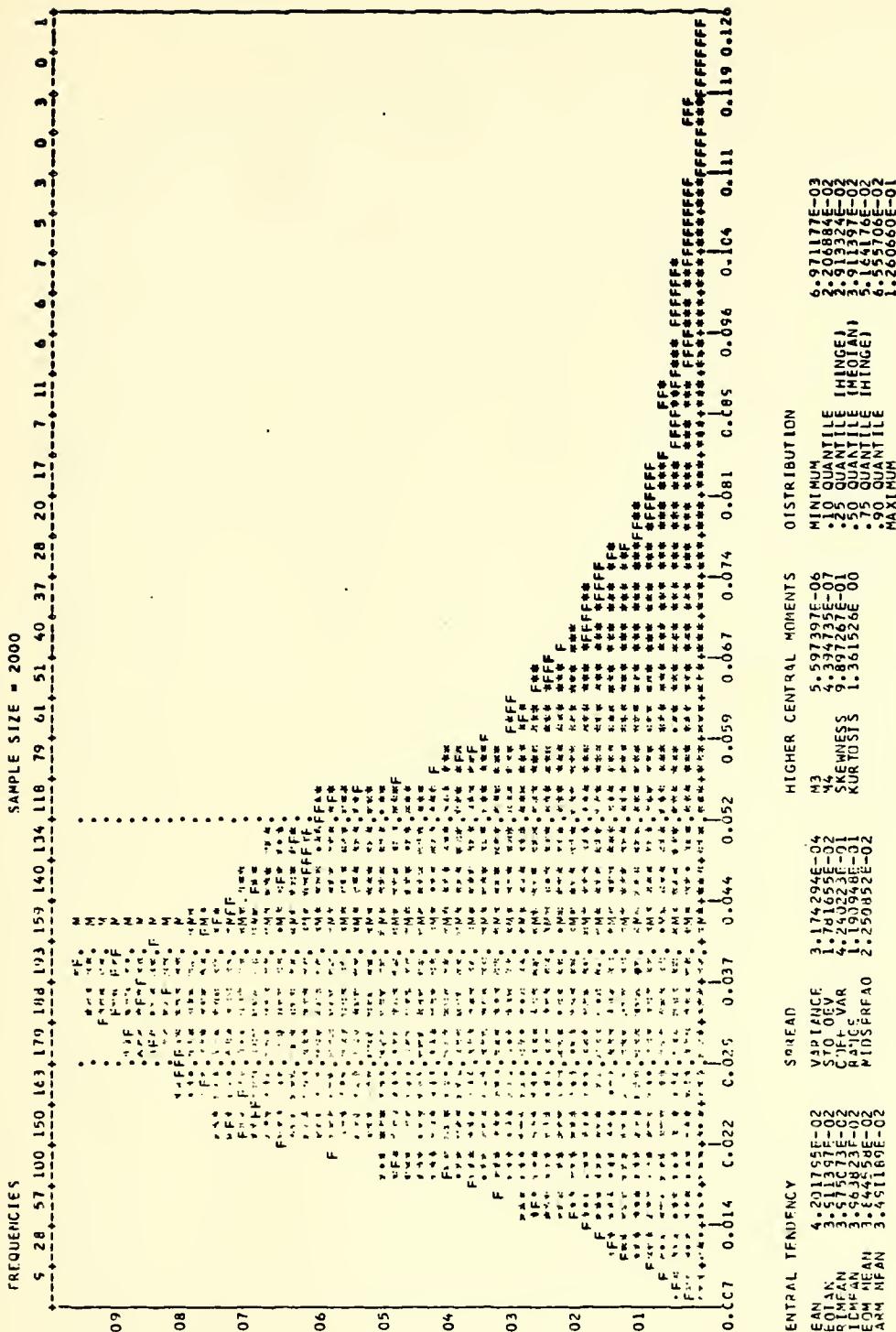


ESTIMATED CENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000      TRIANGULAR WINDOW      RANGE / 10  
 ESTIMATED CENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS



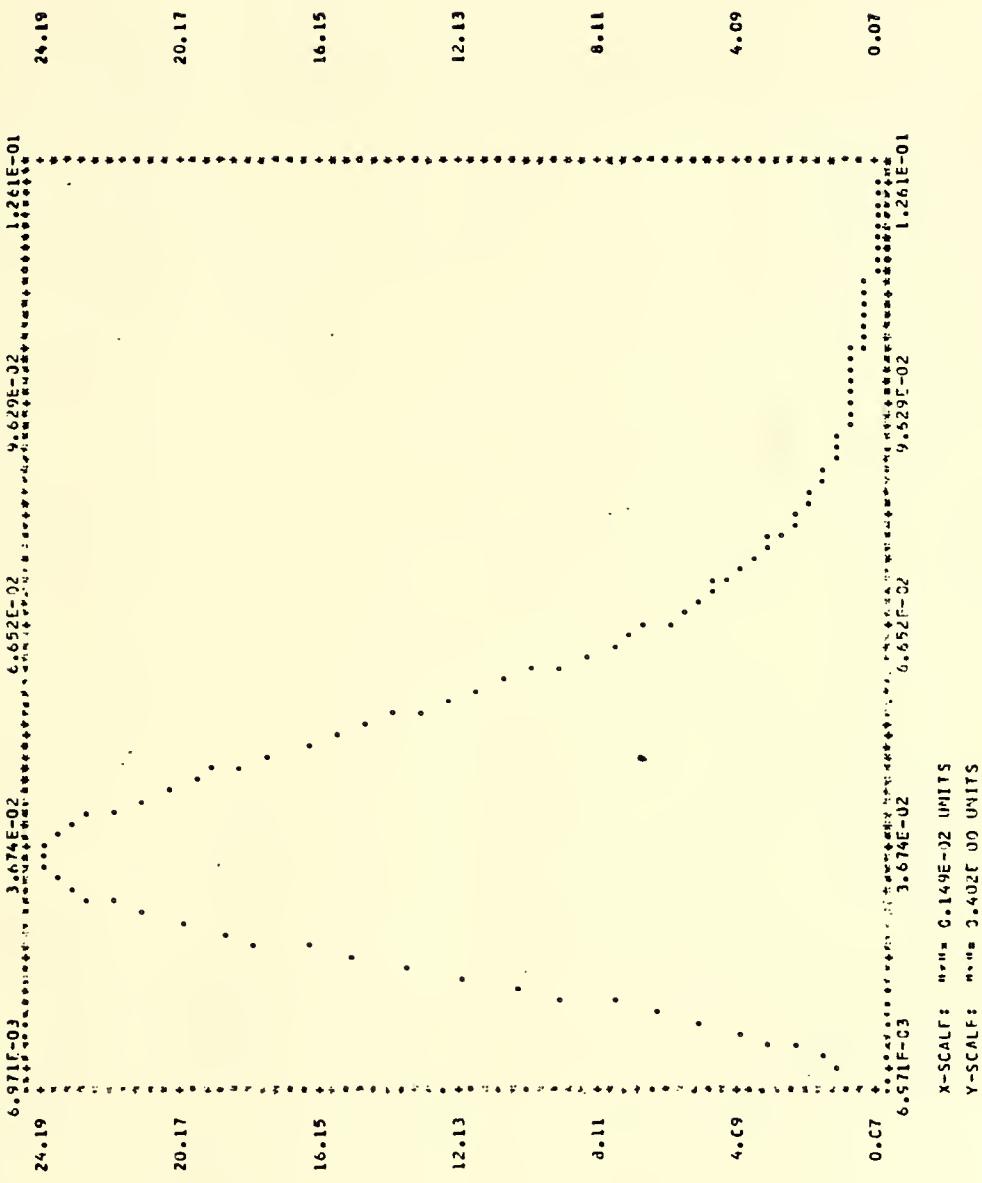






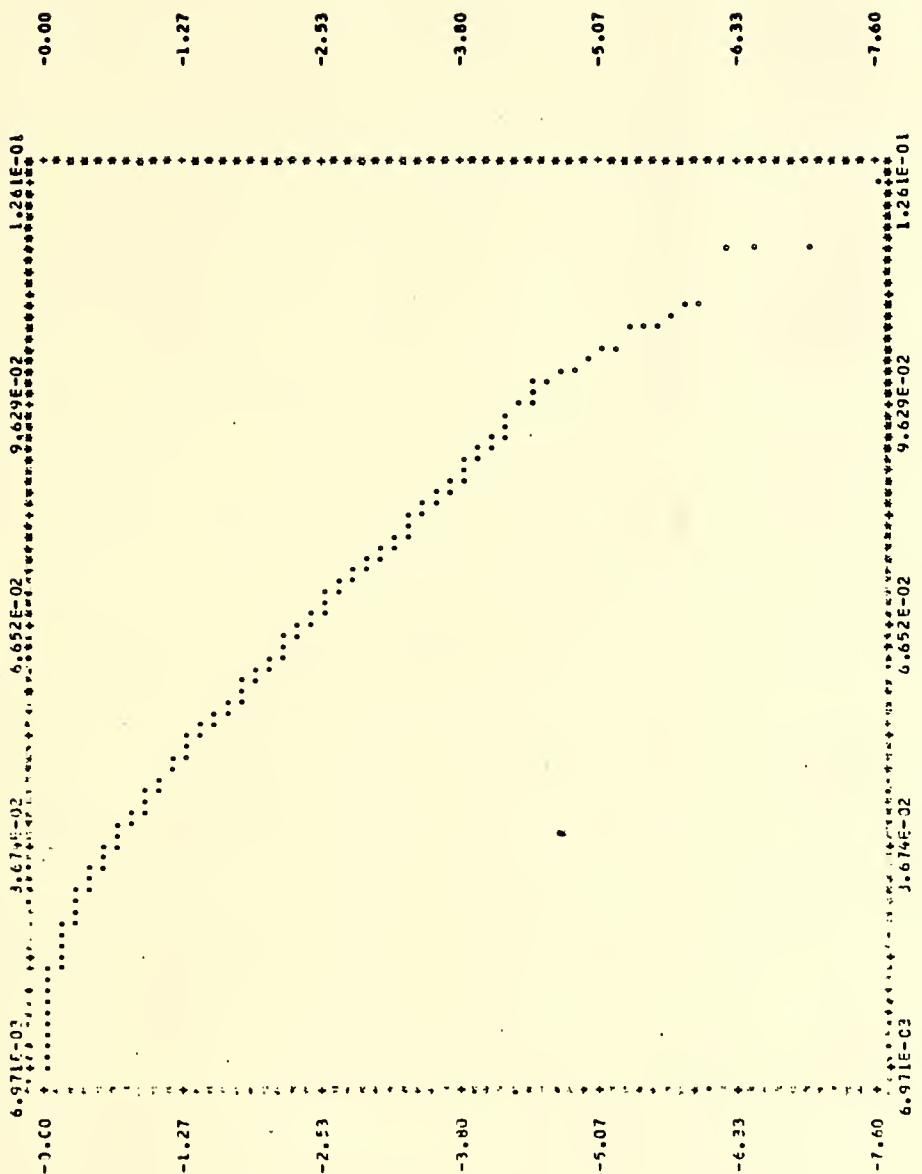
**INTEGRAL SQUARE** **NECPM SAMPLE SIZE**  $M = 2000$   
**UNIFORM RANDOM VARIABLE SAMPLE SIZE**  $N = 200$   
**TRIANGULAR WINDOW.** **PANCHIETH = 1/50ATIN)**



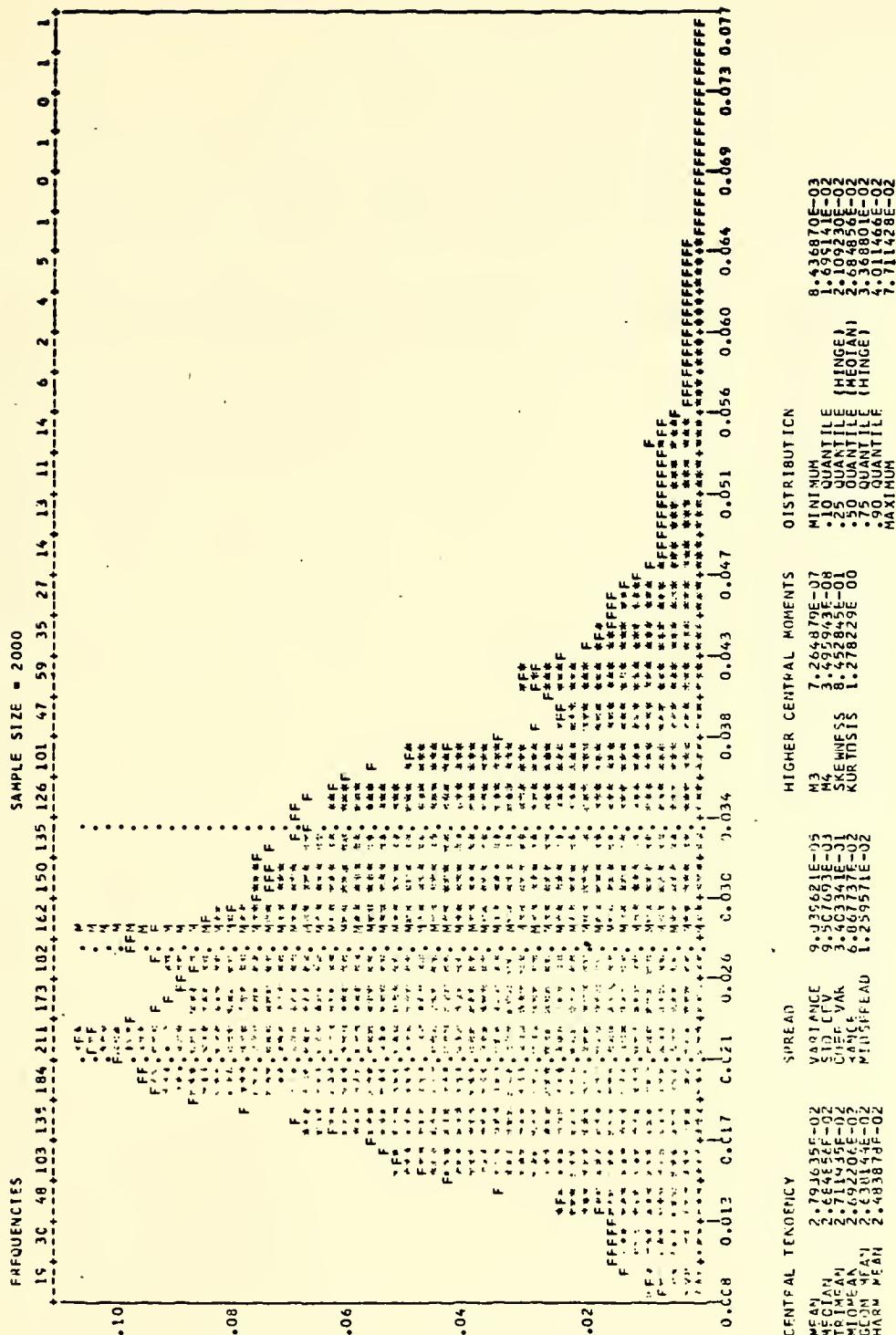


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE TERM RANDOM VARIABLE  
 TRAPEZOIDAL SQUARE NCRM SAMPLE SIZE = 2000 TRIANGULAR NINDW DANDWDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS

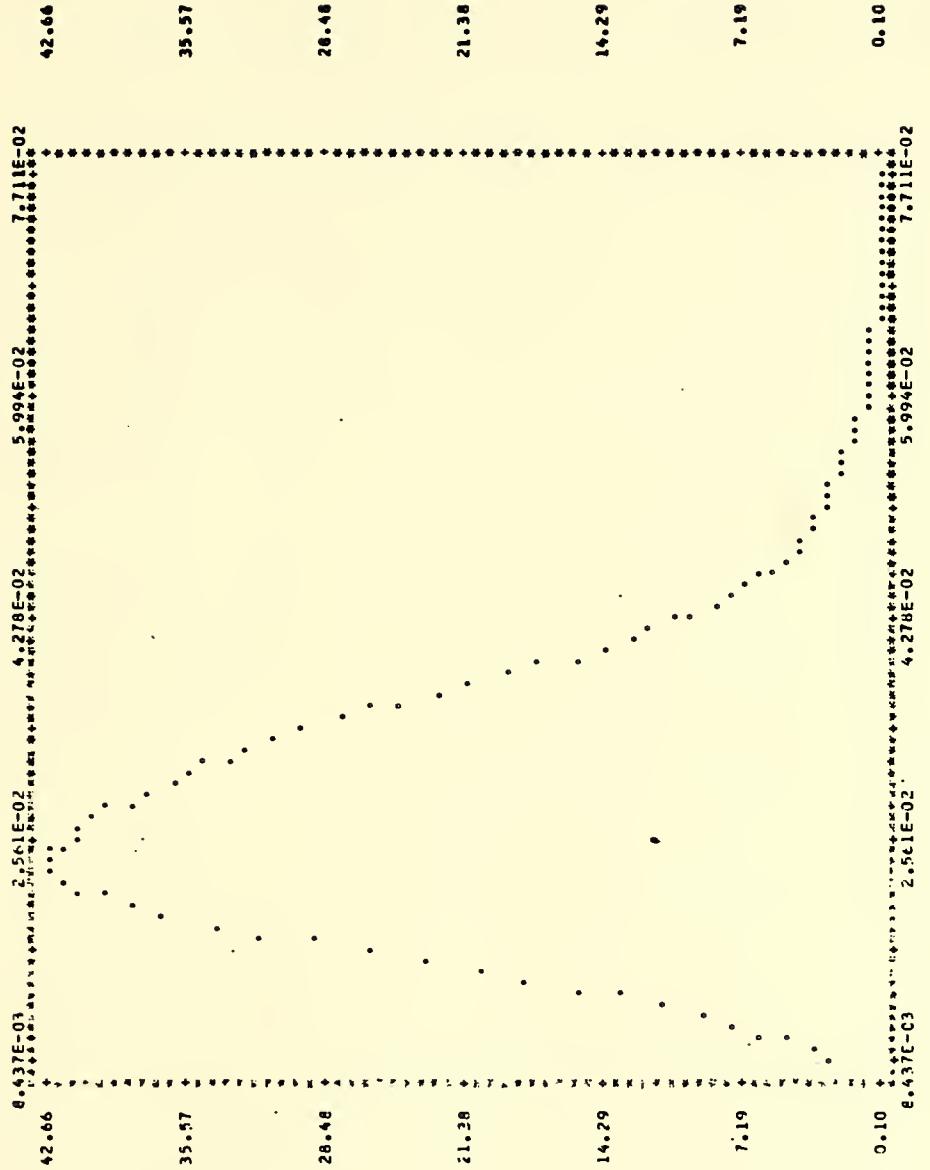








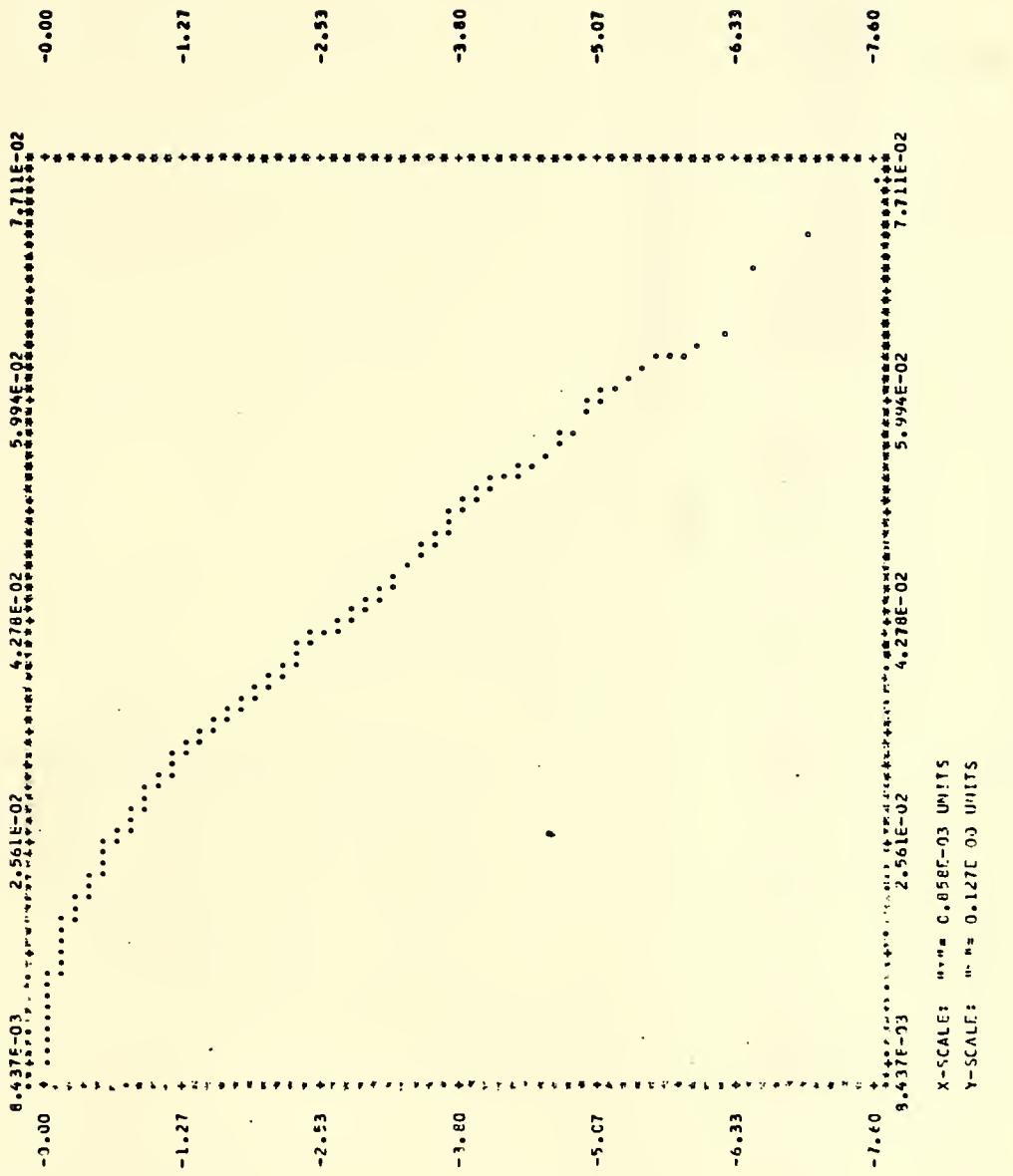




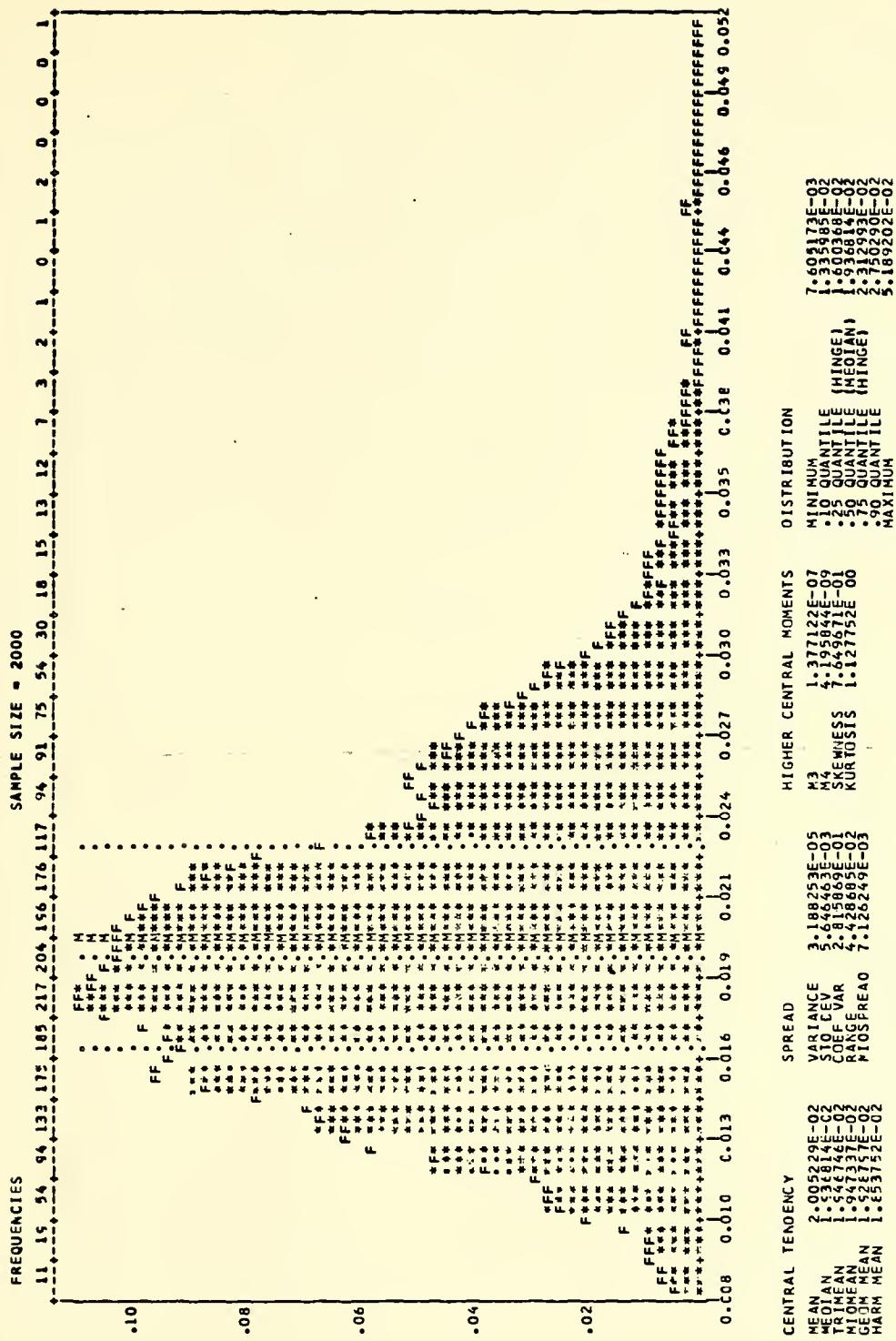
X-SCALE: n.n = C.858E-03 UNITS  
 Y-SCALE: n.n = C.7CSE 00 UNITS

ESTIMATE) DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS



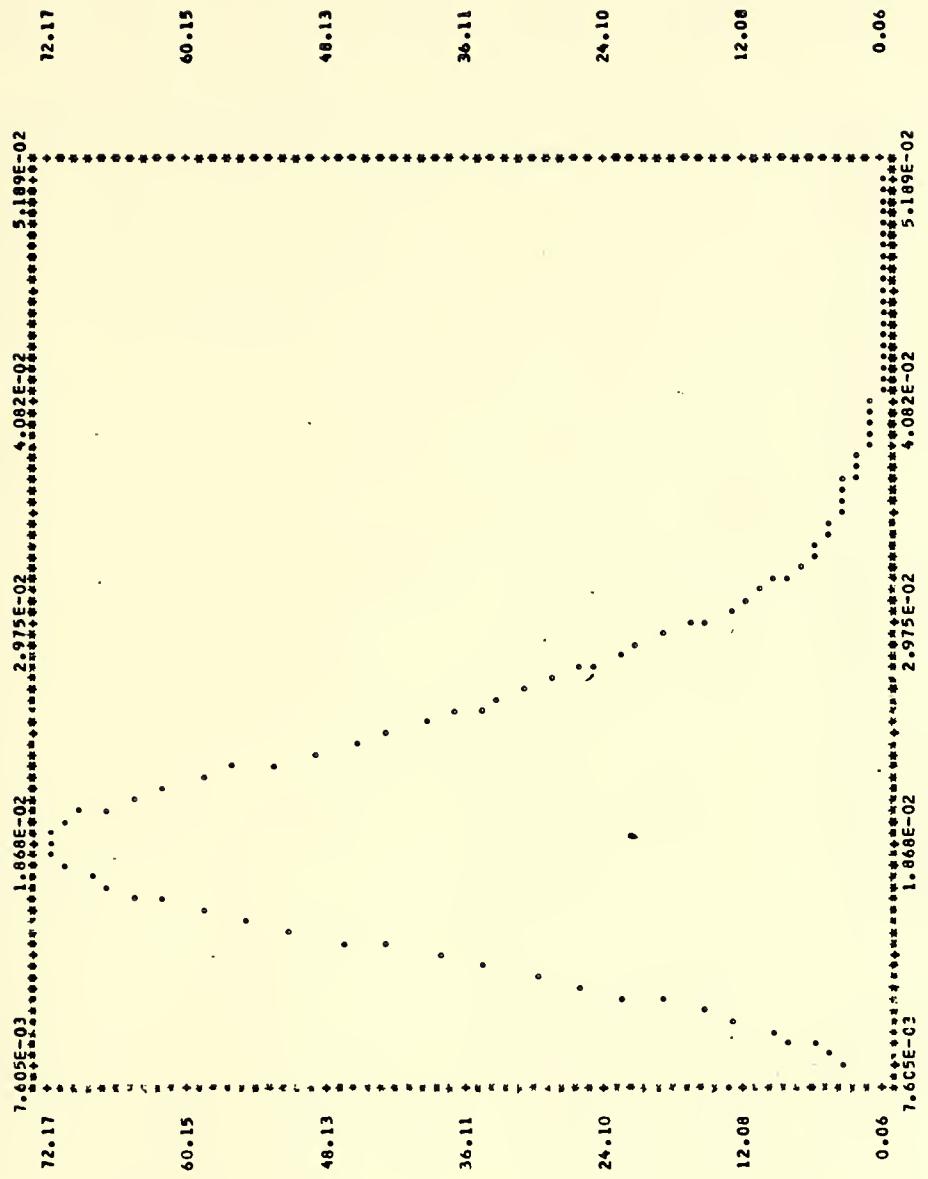






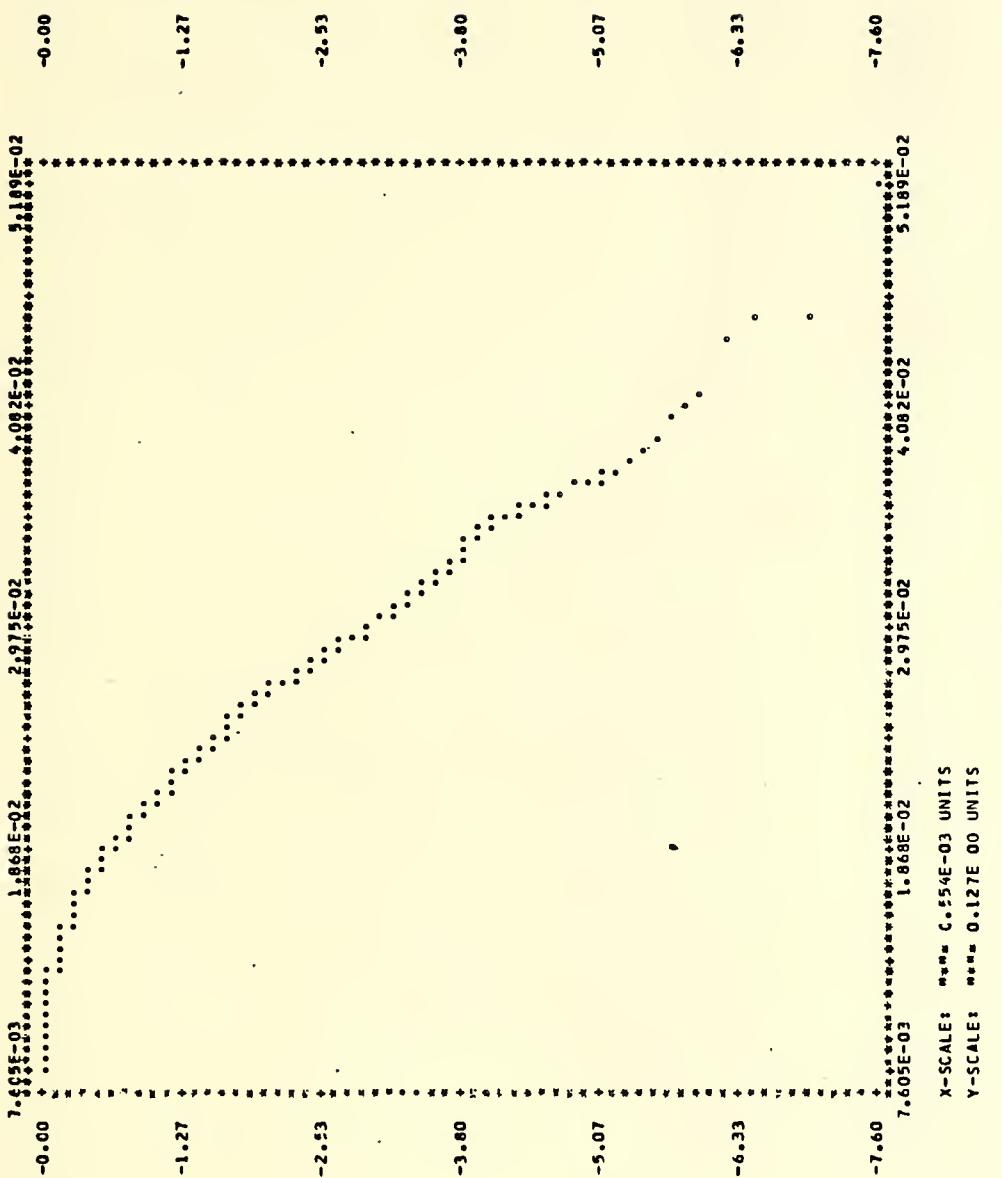
INTEGRAL SQUARE NORM SAMPLE SIZE M = 2000  
UNIFORM RANDOM VARIABLE SAMPLE SIZE N = 1000  
TRIANGULAR WINDOW.    EANWICHT = 1/SQRT(N)



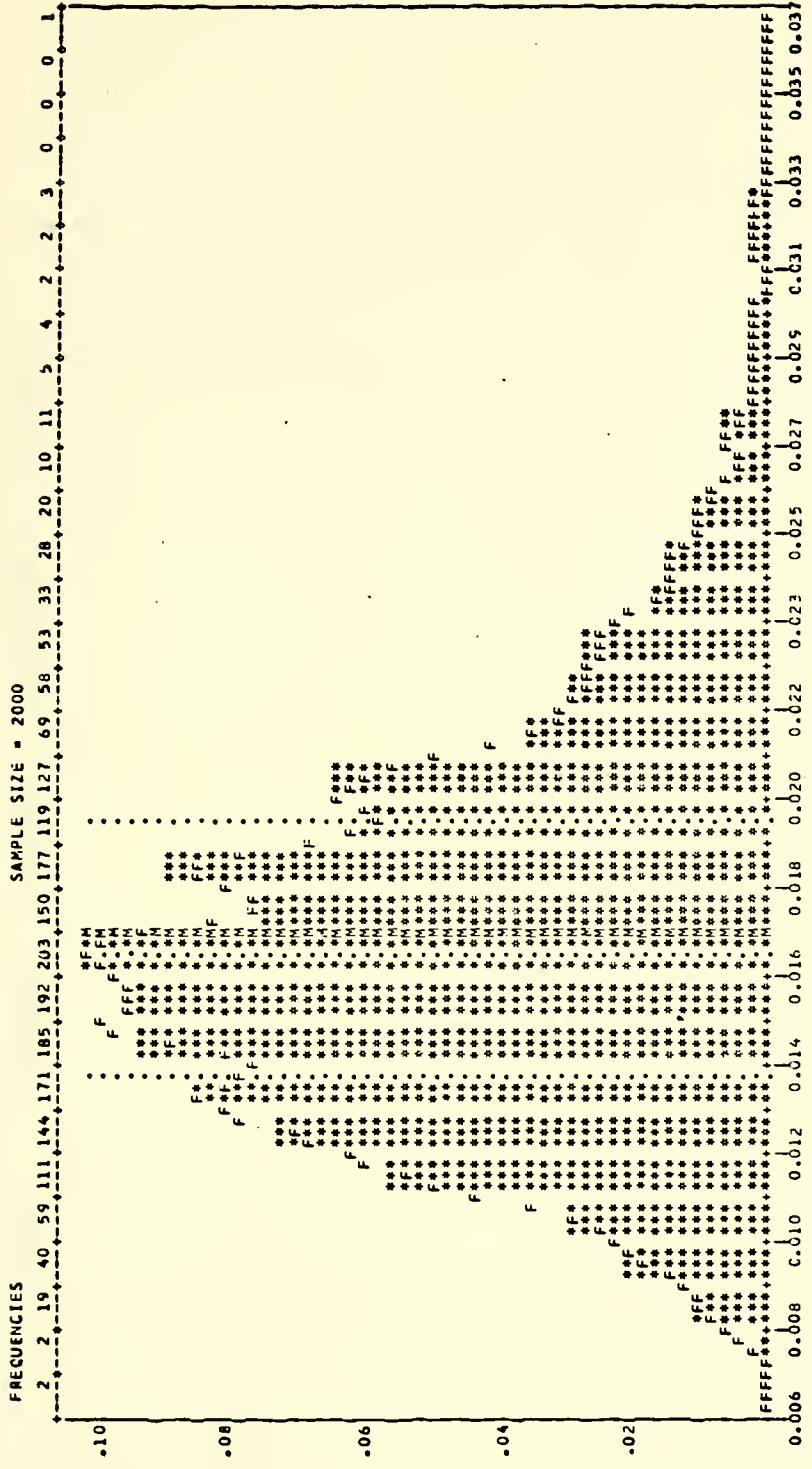


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS









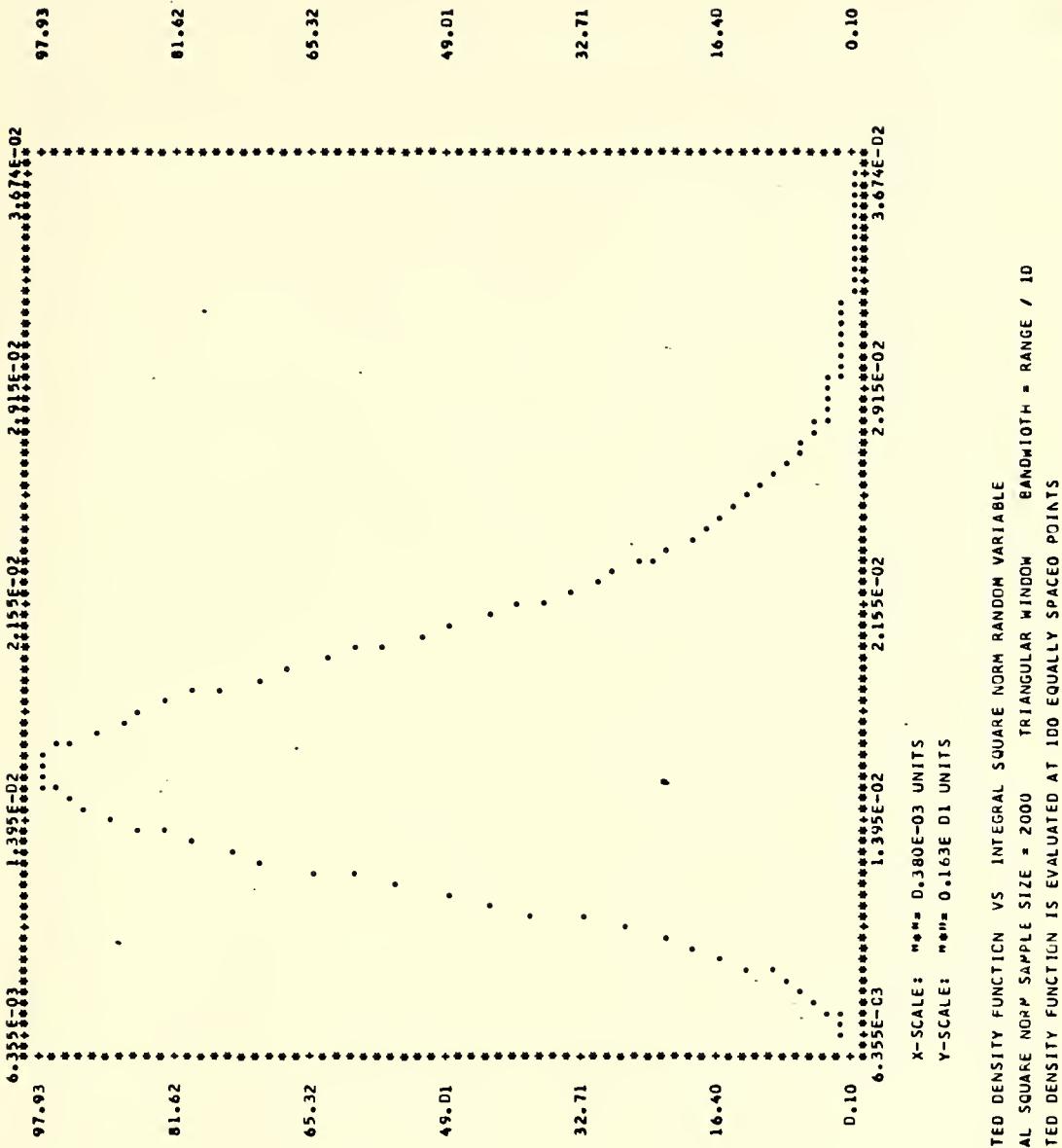
CENTRAL TENDENCY	SPREAD			HIGHER CENTRAL MOMENTS			DISTRIBUTION		
	MEAN	VARIANCE	SD DEV	M <sub>3</sub>	M <sub>4</sub>	SKEWNESS	MINIMUM	MAXIMUM	
MEAN	1.662469E-02	1.633689E-05	4.21656E-08	1.458314E-02	9.75644E-10	6.239231E-01	1.458314E-02	1.368024E-02	
MEAN	1.64629E-02	4.041462E-03	2.43684E-02	3.038736E-01	6.237561E-01	7.2 QUANTILE (MINGI)	1.64629E-02	1.36326E-02	
TRIMMED MEAN	1.629606E-02	4.038684E-02	2.435884E-02	3.037561E-01	6.237561E-01	7.2 QUANTILE (MINGI)	1.629606E-02	1.36326E-02	
TRIMMED MEAN	1.624994E-02	5.335086E-03	2.434994E-02	3.036436E-01	6.236436E-01	7.2 QUANTILE (MINGI)	1.624994E-02	1.36257E-02	
GEOM. MEAN	1.5671826E-02	1.508666E-02	1.508666E-02	1.508666E-01	1.508666E-01	1.508666E-01	1.508666E-01	1.508666E-01	
HARM. MEAN	1.5671826E-02	1.508666E-02	1.508666E-02	1.508666E-01	1.508666E-01	1.508666E-01	1.508666E-01	1.508666E-01	

INTEGRAL SQUARE NORM SAMPLE SIZE  $N = 2000$

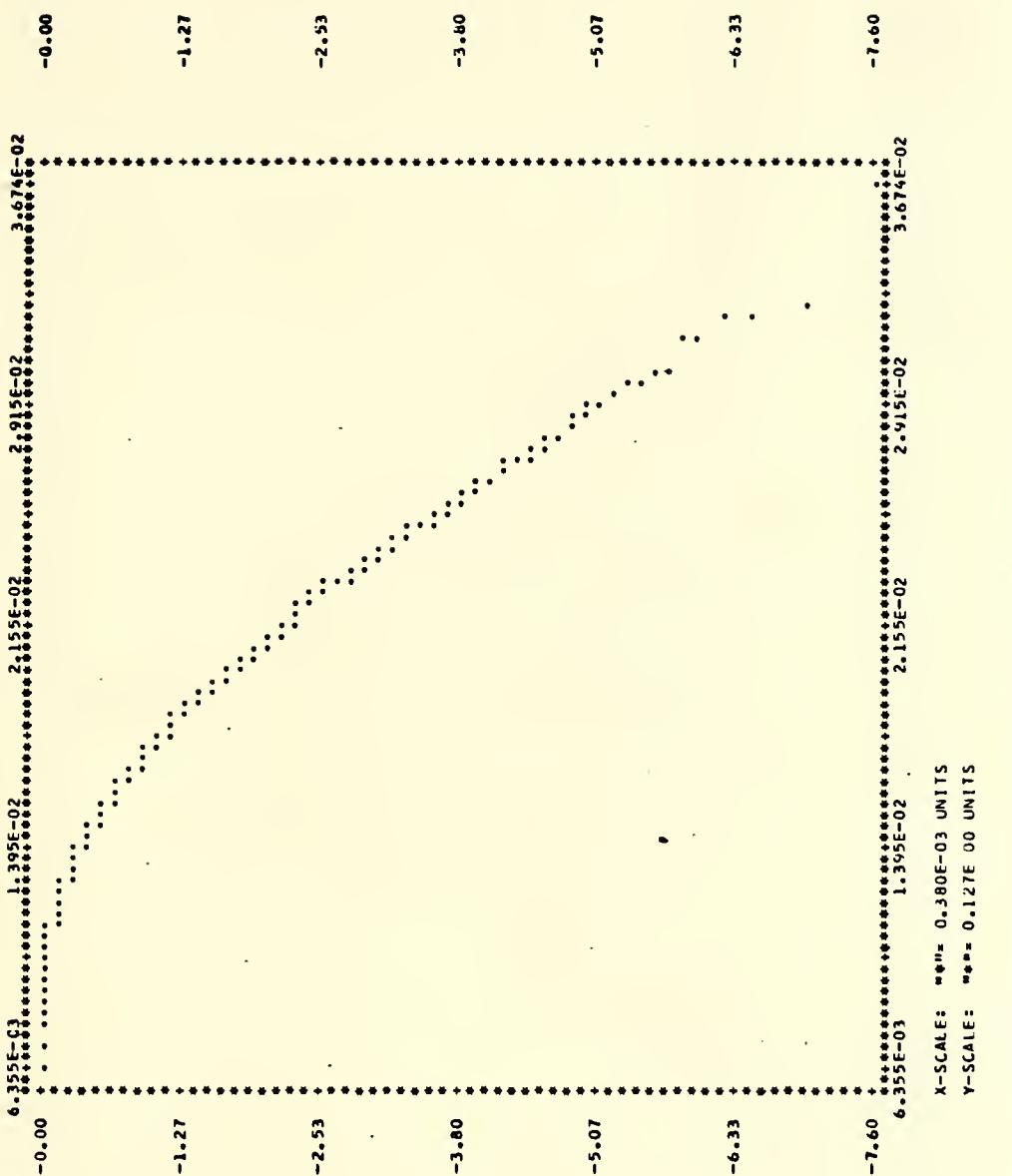
UNIFORM RANDOM VARIABLE SAMPLE SIZE  $N = 1500$

TRIANGULAR WINDOW =  $1/\sqrt{N}$

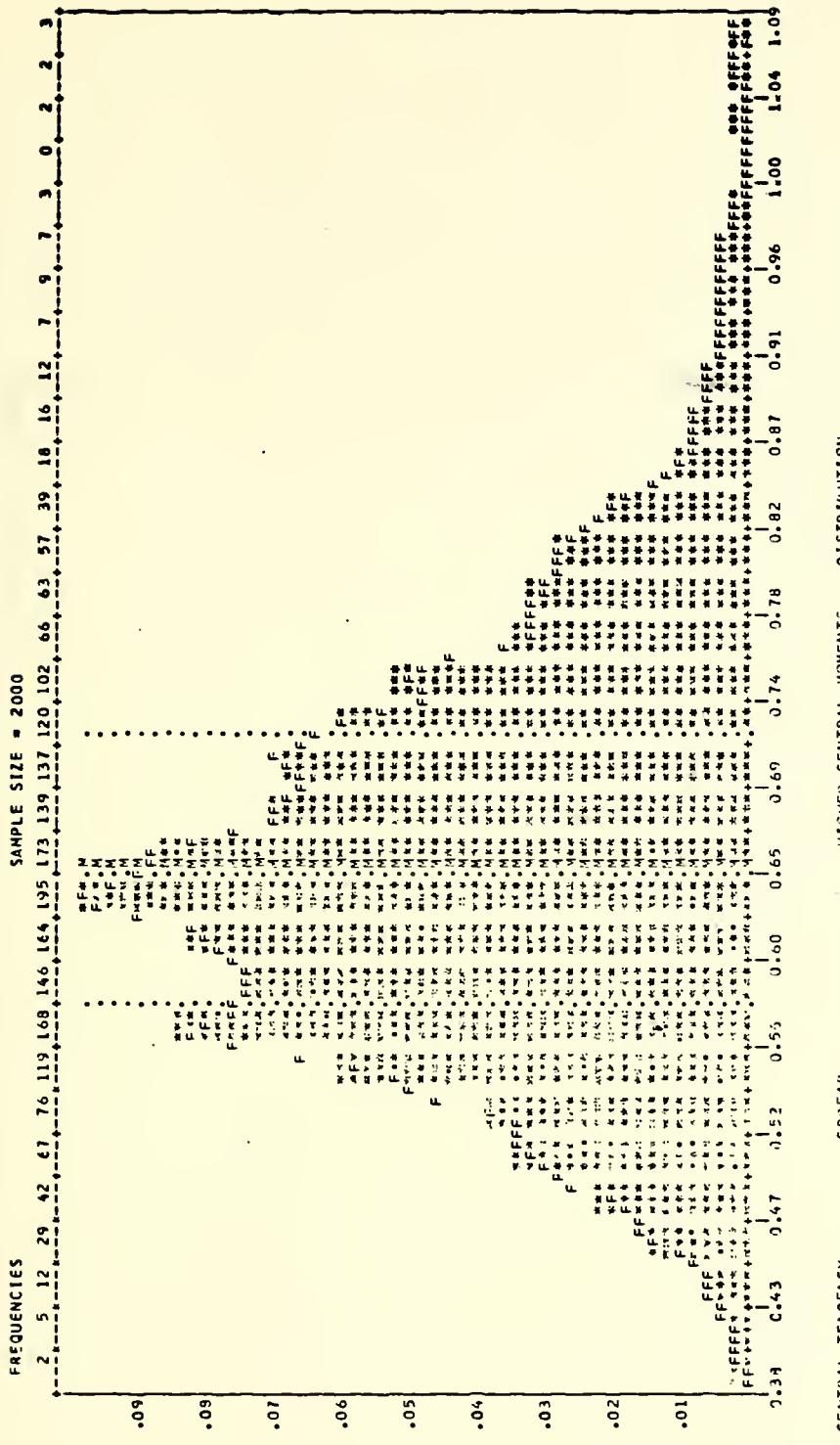




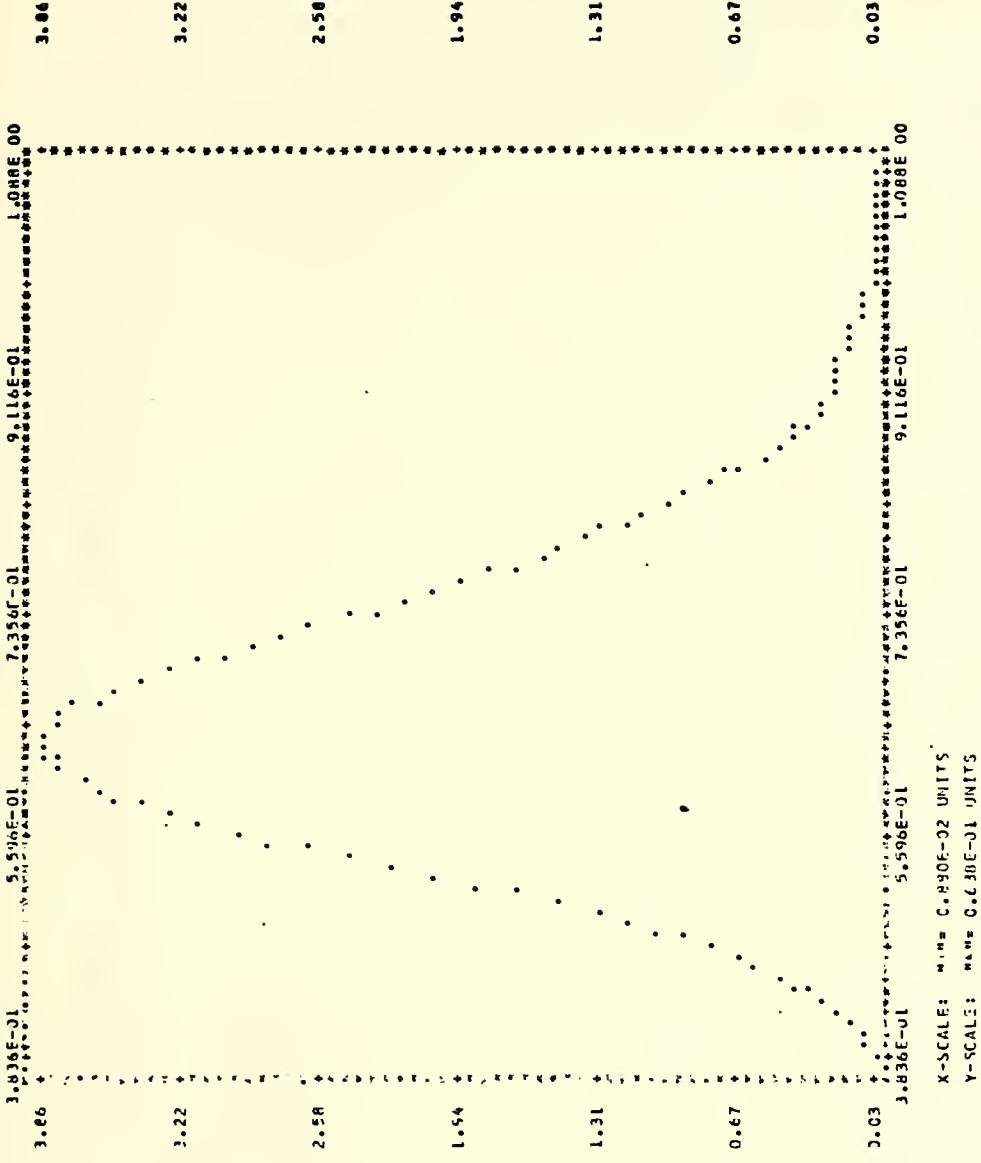






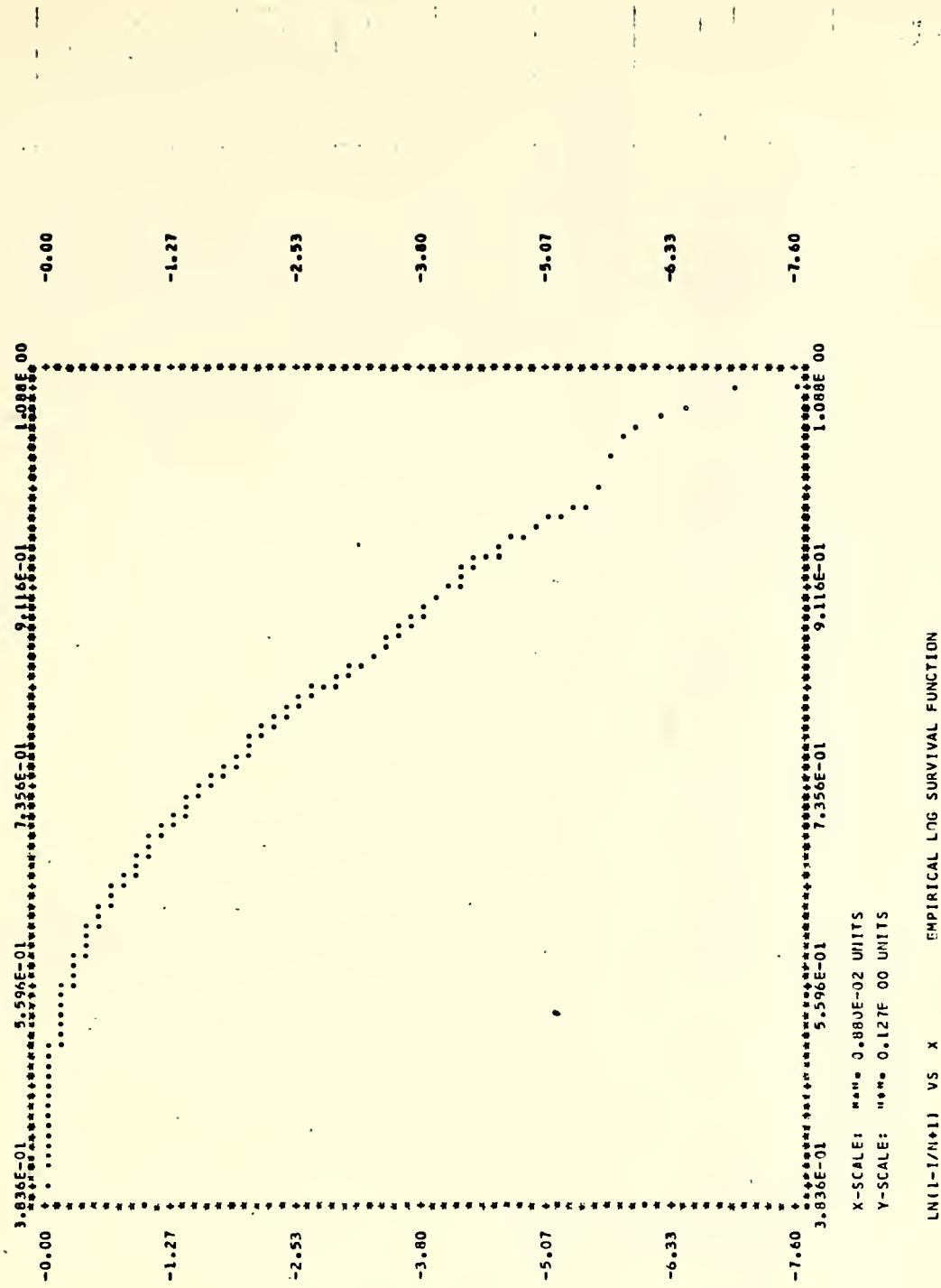




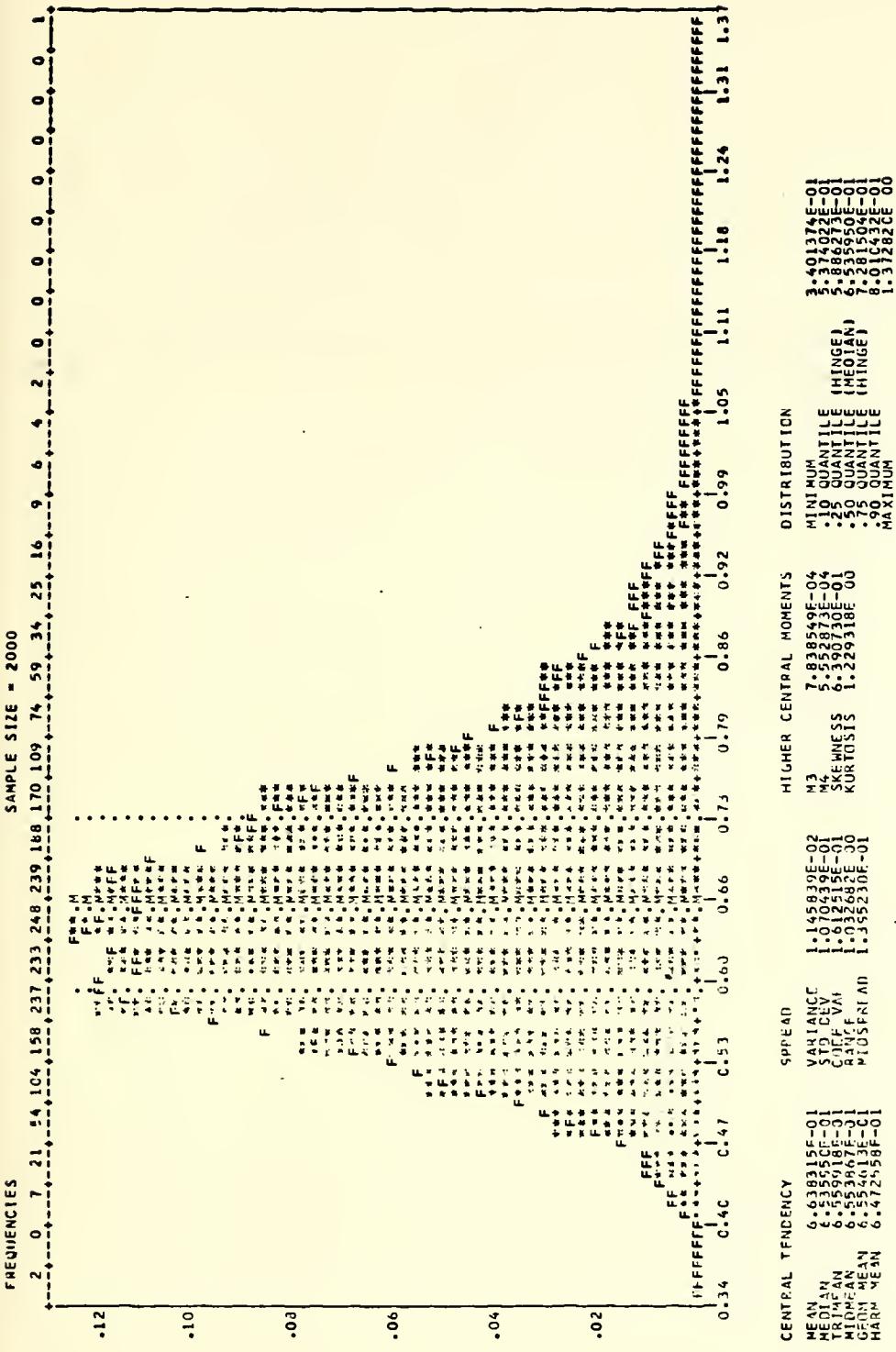


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTERVAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS



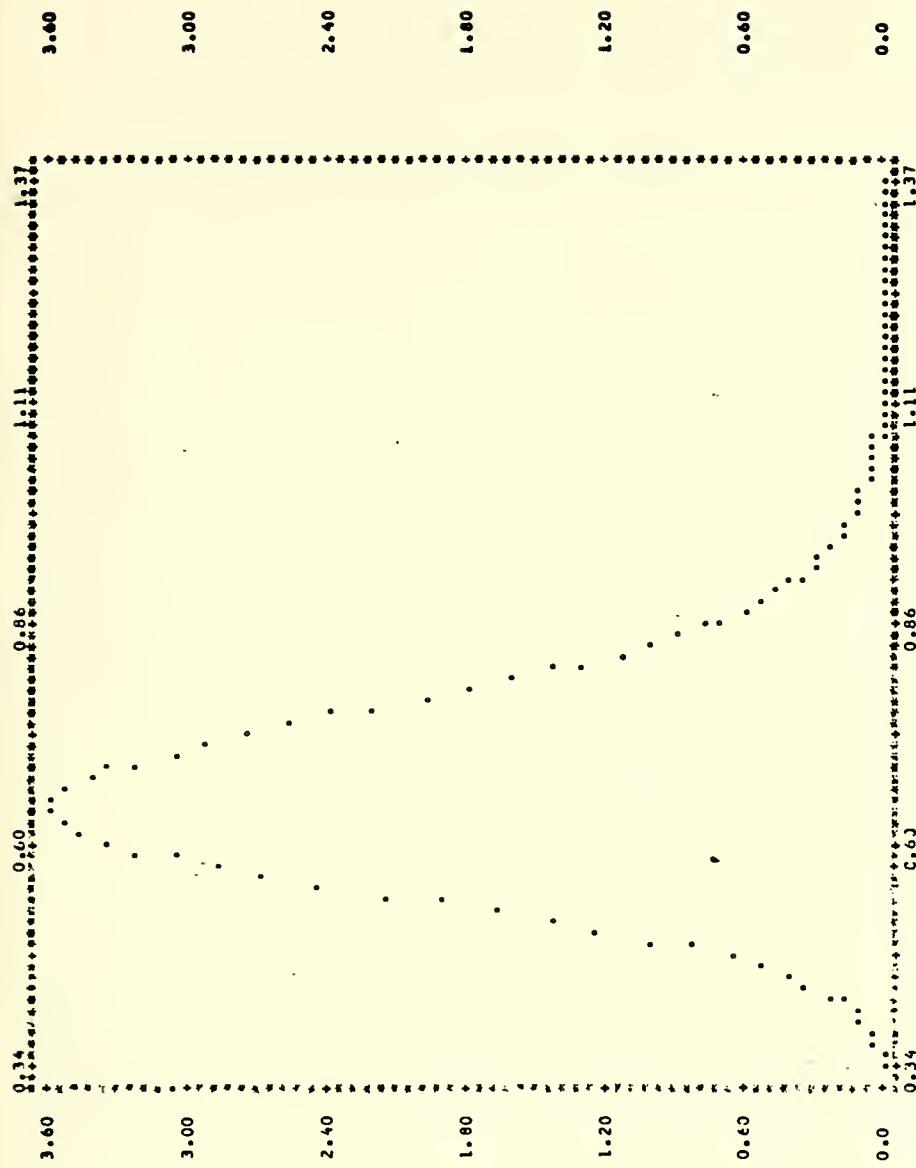






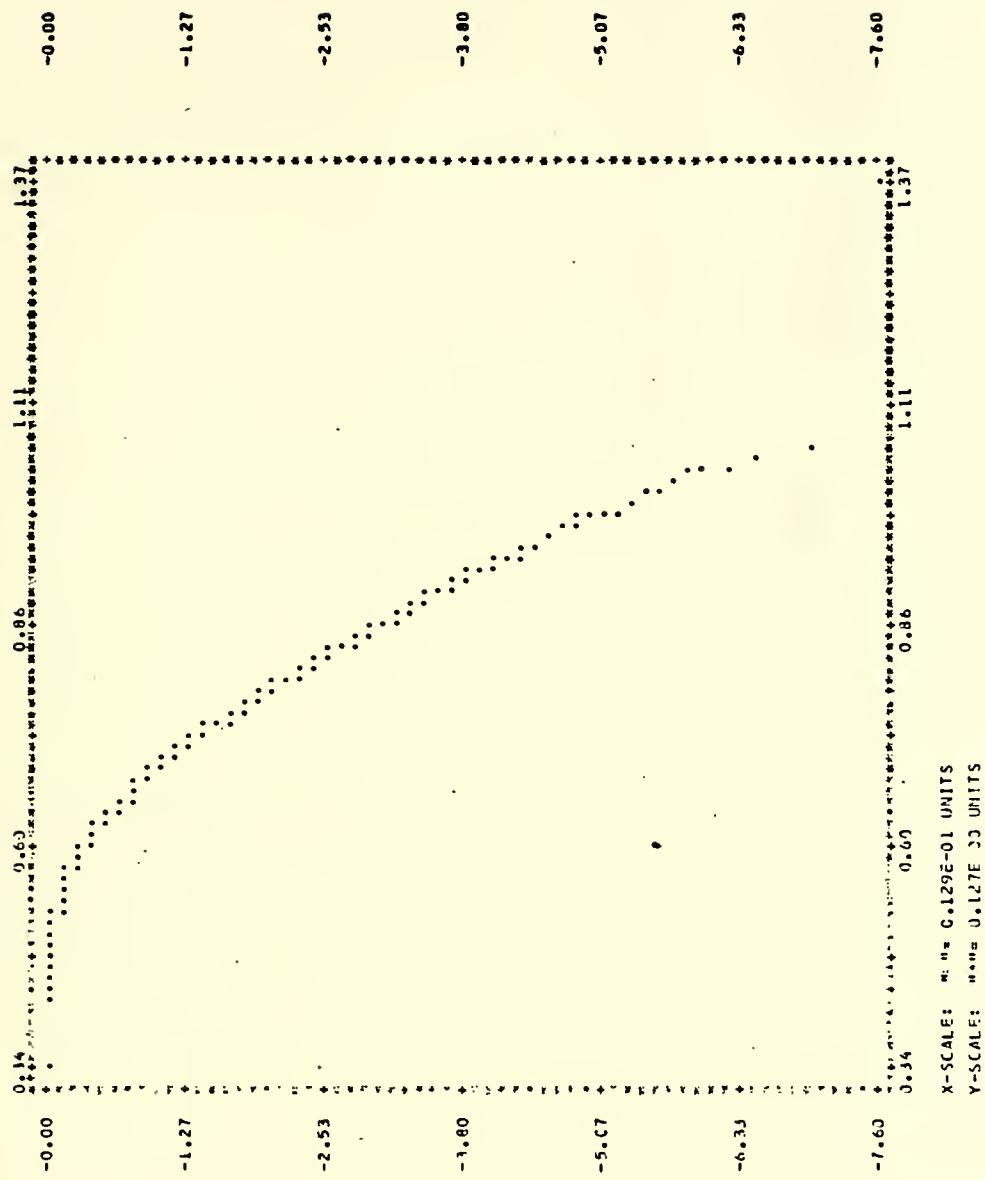
INTEGRAL SQUARE MTRV SAMPLE SIZE M = 2000  
 UNIFORM RANGE VARIABLE SAMPLE SIZE N = 2000  
 TRIANGULAR DISTRIBUTION FRANCHISES = 1/N



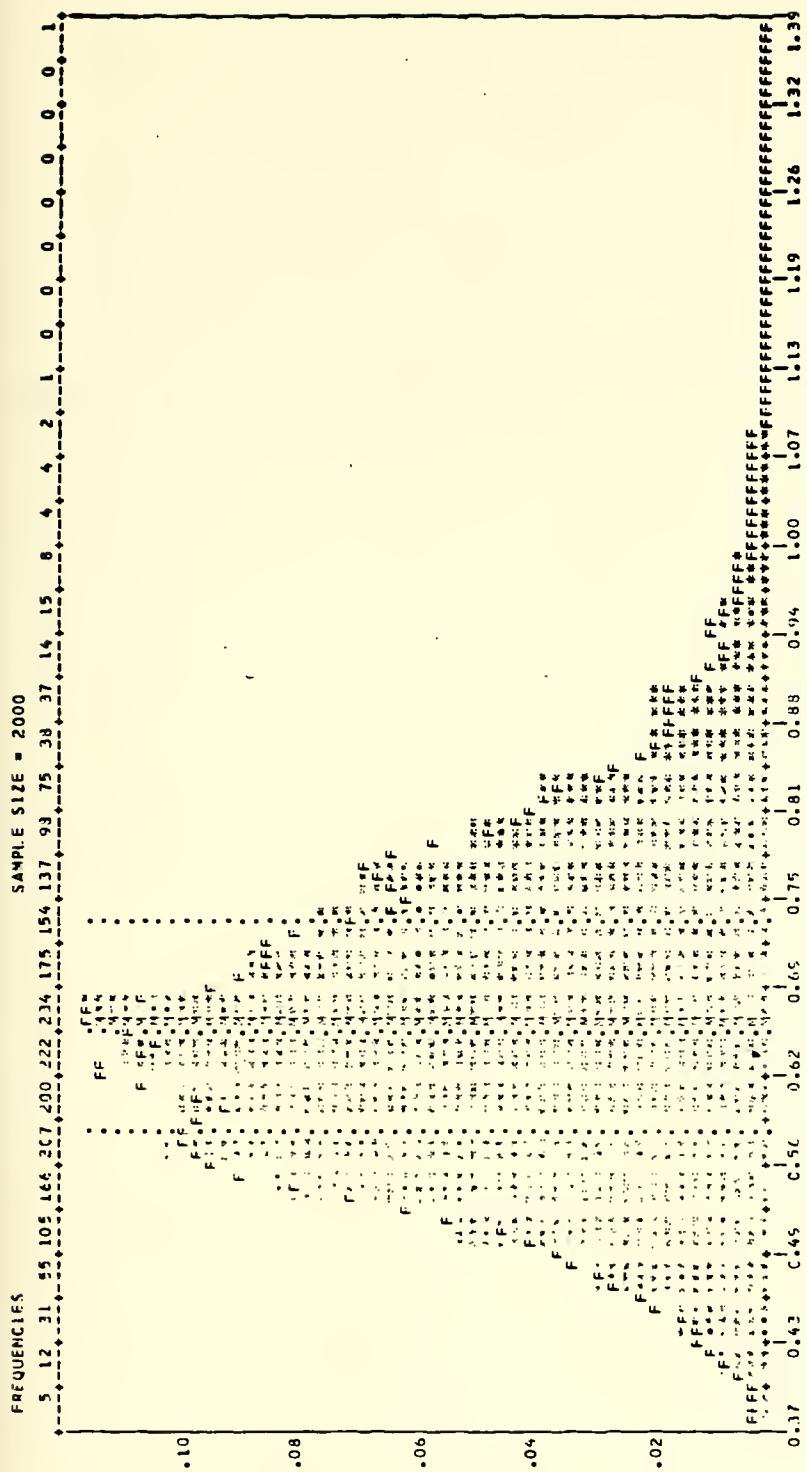


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000    TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS





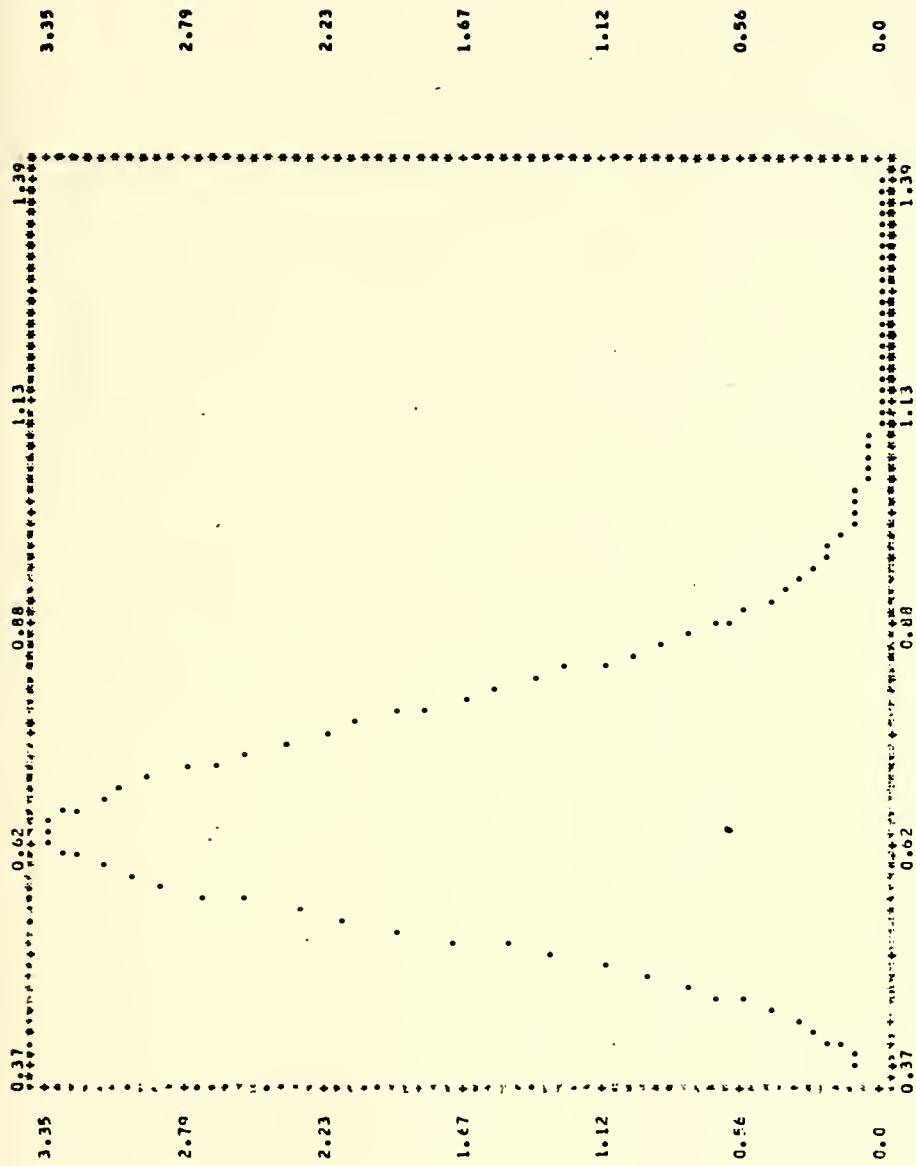




CENTRAL TENDENCY	SPRT (n)	HIGHER CENTRAL MOMENTS				DISTRIBUTION
		M3	M4	SKEWNESS	KURTOSIS	
MEAN	6.51197E-01	1.32616E-32	0.74442E-04	0.10000E+00	0.74442E-04	MINIMUM
MECHANICAL MEAN	6.52858E-01	STD	0.748285E-01	0.75000E+00	0.748285E-01	0.10 QUANTILE
TRIMMED MEAN	6.53087E-01	CDF	0.74644E-01	0.75000E+00	0.74644E-01	0.25 QUANTILE
TRIMMED MEAN (10%)	6.52624E-01	PDF	0.74686E-01	0.75000E+00	0.74686E-01	0.50 QUANTILE
GEOMETRIC MEAN	6.51414E-01	HIGHWAY	0.538466E-10	0.90000E+00	0.538466E-10	0.75 QUANTILE
HARM. MEAN	6.41770E-01	TAIPEI	0.538466E-10	0.90000E+00	0.538466E-10	0.90 QUANTILE
						MAXIMUM

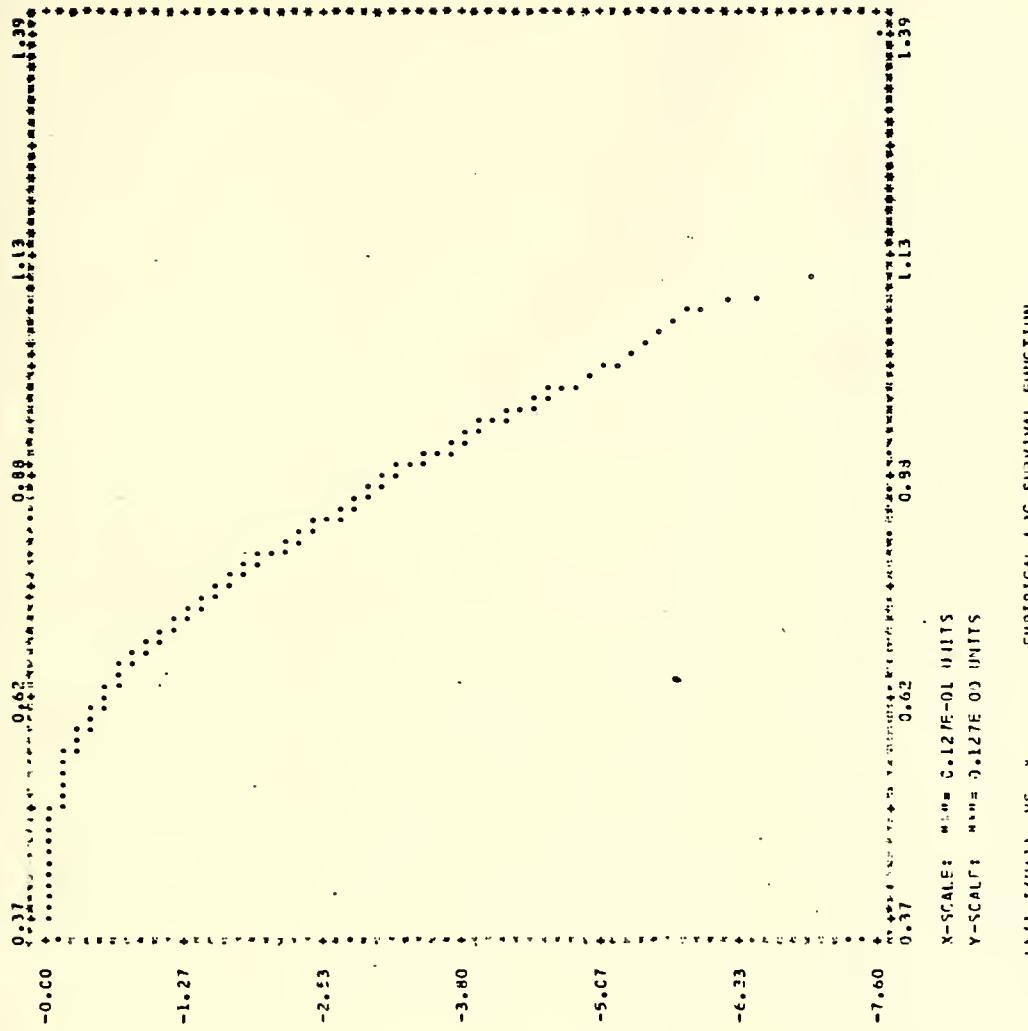
INTEGRAL SQUARE \* (THE SAMPLE SIZE  $M = 200$ )  
 UNIFORM RANDOM VARIABLE SAMPLE SIZE  $N = 200$   
 TRIANGULAR DISTRIBUTION SAMPLE SIZE  $N = 1/N$



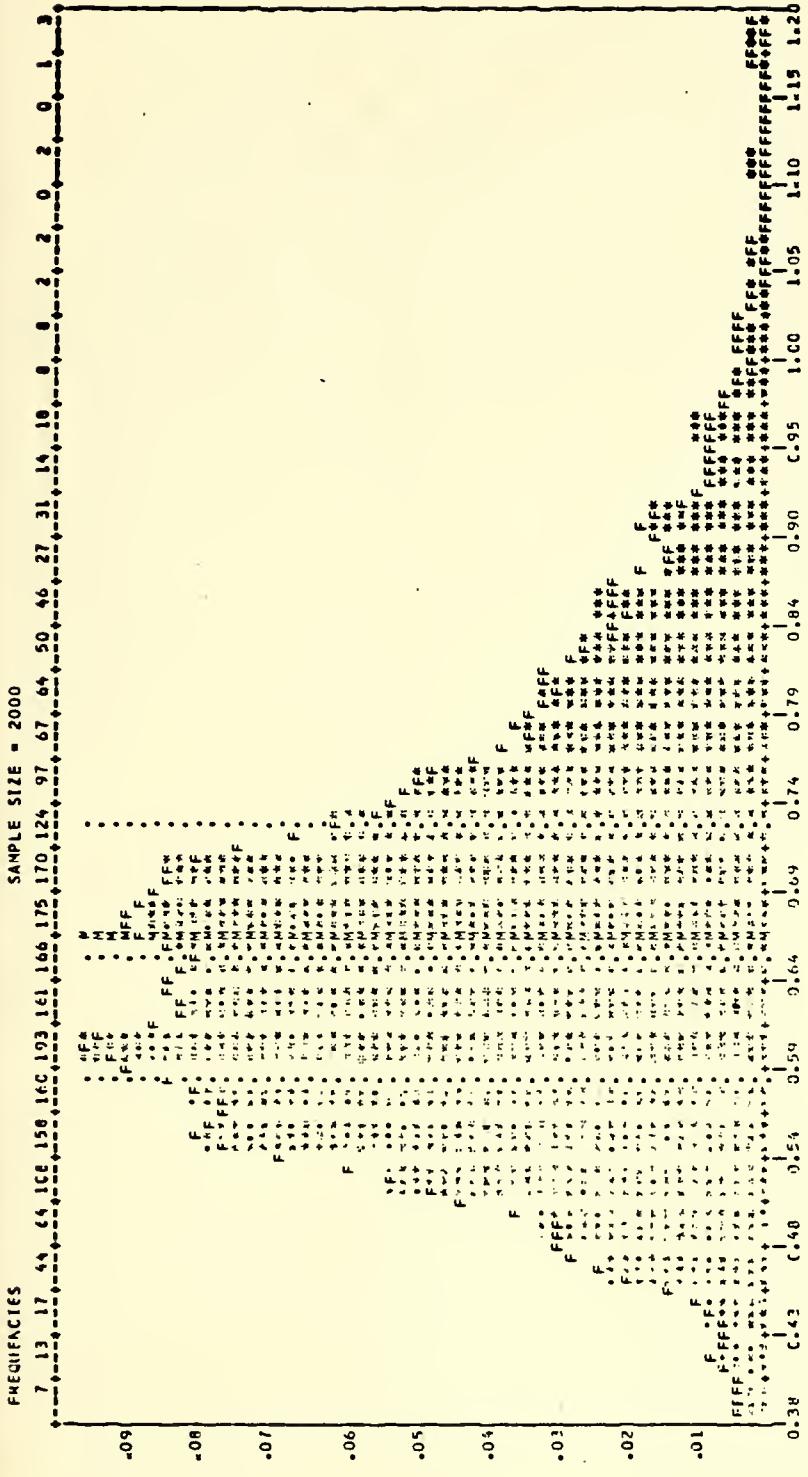


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE ACM RANDOM VARIABLE  
 INTEGRAL SQUARE INCM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS







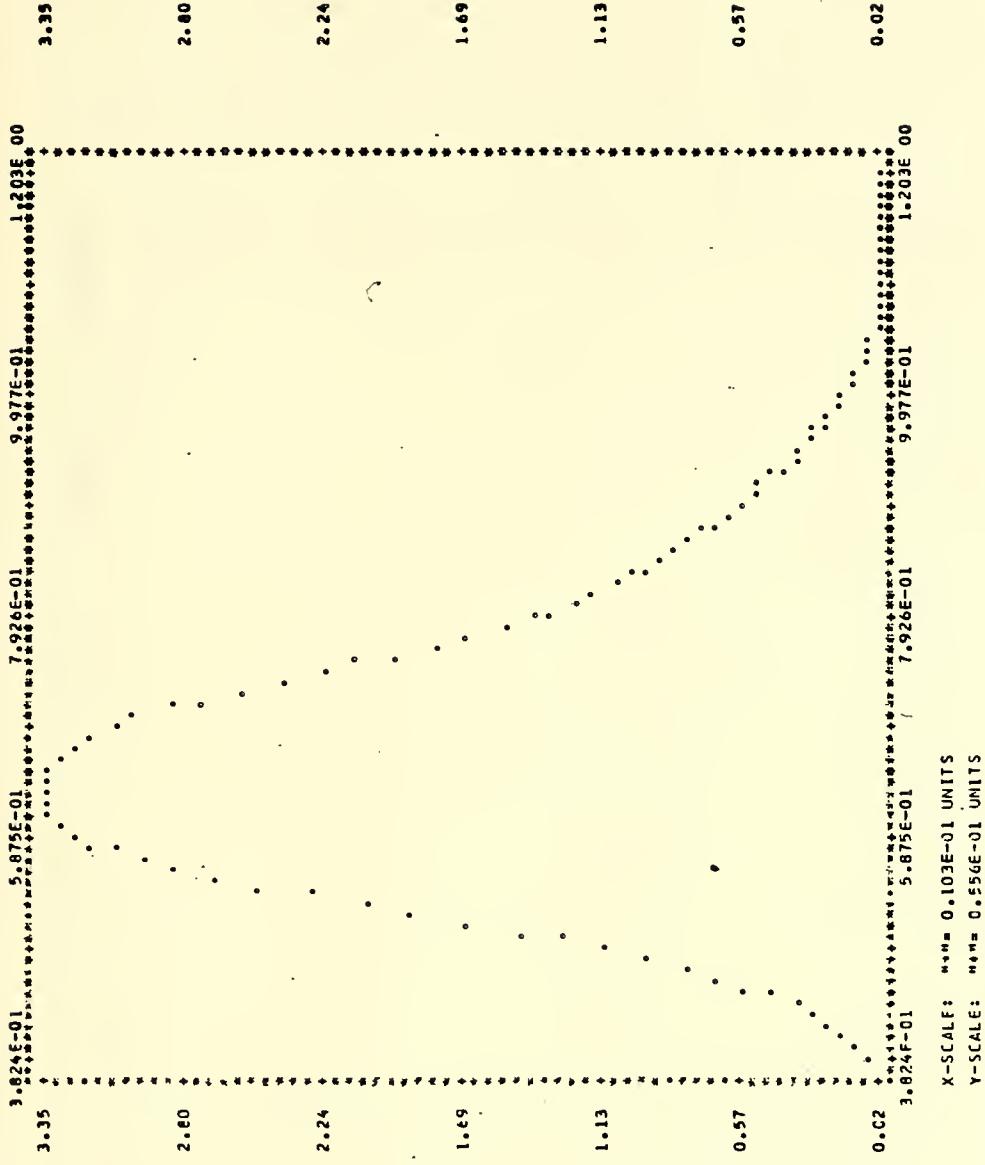


CENTRAL TENDENCY	MEAN	MEDIAN	MODE	STANDARD DEVIATION
MEAN	6.4111	6.4111	6.4111	0.205
MEDIAN	6.4111	6.4111	6.4111	0.205
MODE	6.4111	6.4111	6.4111	0.205
STANDARD DEVIATION	0.205	0.205	0.205	0.205

HIGHER CENTRAL MOMENTS	DISTRIBUTION
M <sub>3</sub>	1.179938E-03
M <sub>4</sub>	7.449632E-04
SKEWNESS	0.369673E-01
KURTOSIS	6.92111E-01

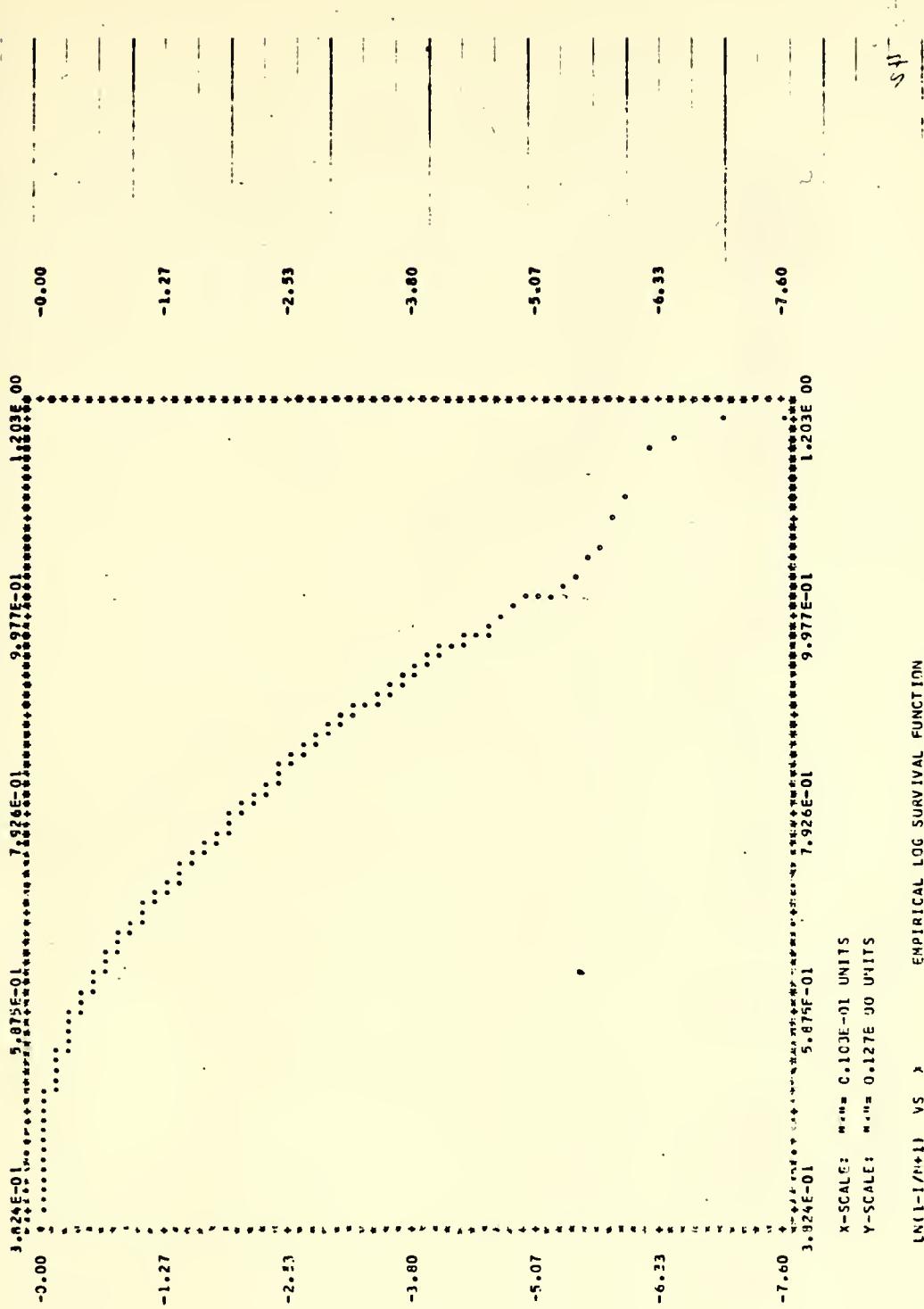
INTERAL SQUARE KEY SAMPLE SIZE  $M = 2000$   
 UNIT IF PANTRY VARIABLE SAMPLE SIZE  $N = 1000$   
 TRIANGULAR WINDOW. LENGTH =  $1/\sqrt{h}$



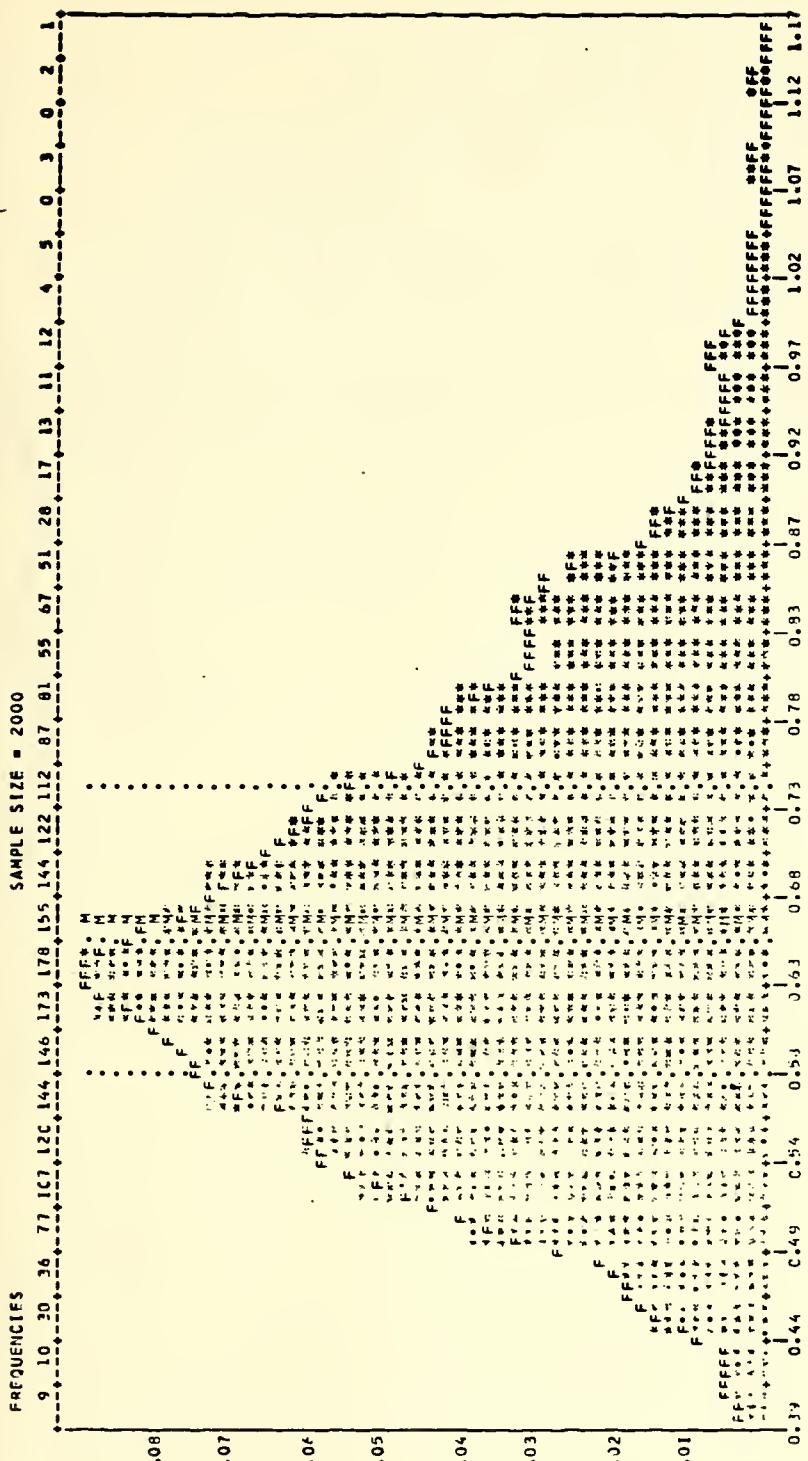


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NCRM RANDOM VARIABLE  
 INTEGRAL SQUAFF NCRM SAMPLE SIZE = 2000      TRIANGULAR WINDOW      BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS





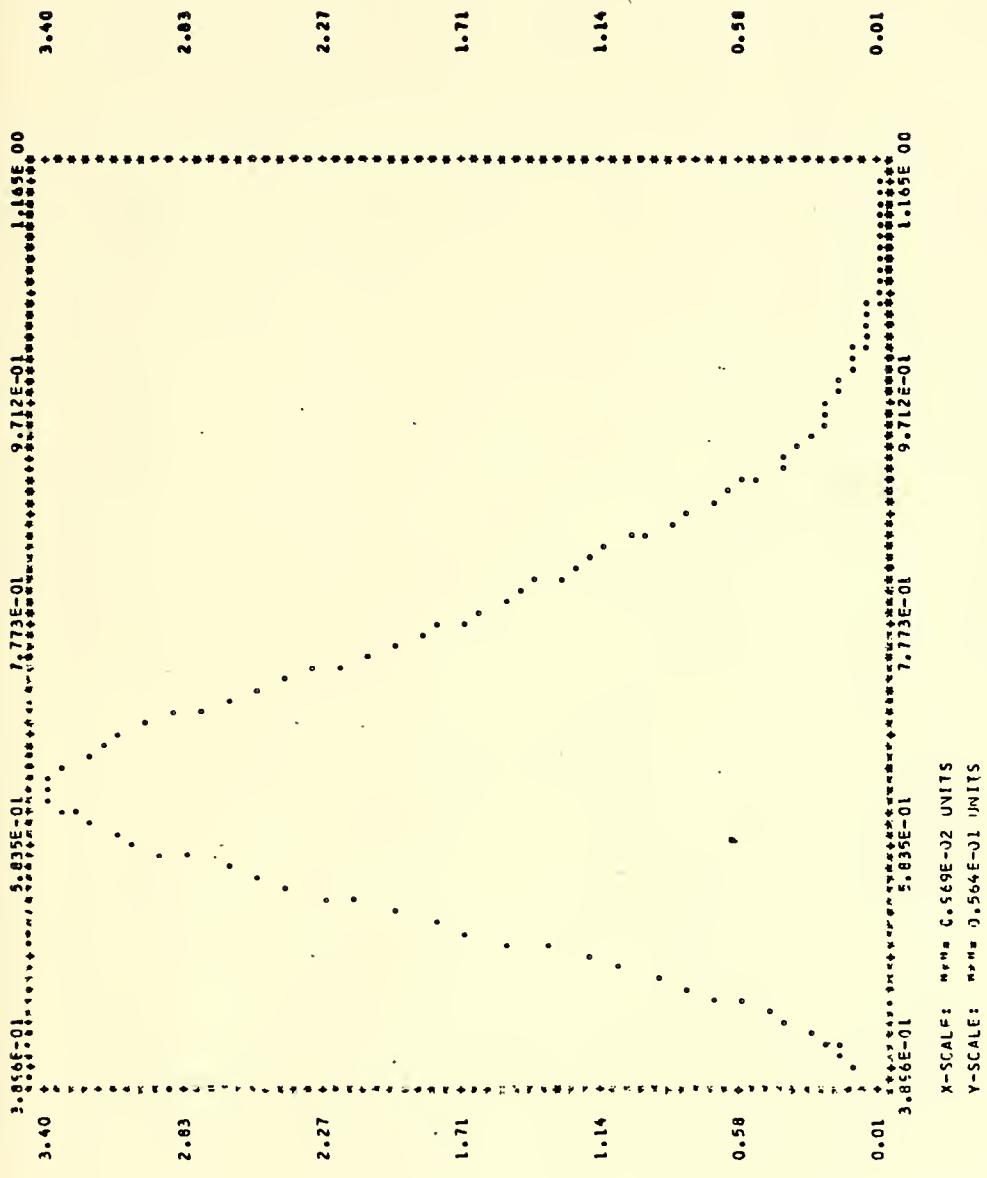




CENTRAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS	DISTRIBUTION
MEAN	$6.644311E-01$	$M_3 = 8.886522E-04$	MINIMUM
UNIFORM	$6.644311E-01$	$M_4 = 0.80736E-04$	0.1 QUANTILE
TRIANGULAR	$6.644311E-01$	$SKEWNESS = 2.424258E-01$	0.25 QUANTILE (MIDGE)
MIDGEE	$6.644311E-01$	$KURTOSIS = 2.692688E-01$	0.5 QUANTILE (MIDGE)
GEOMETRIC	$6.644311E-01$	$M_5 = 0.433347E-01$	0.9 QUANTILE (MIDGE)
HARK	$6.644311E-01$	$M_6 = 0.438082E-01$	MAXIMUM

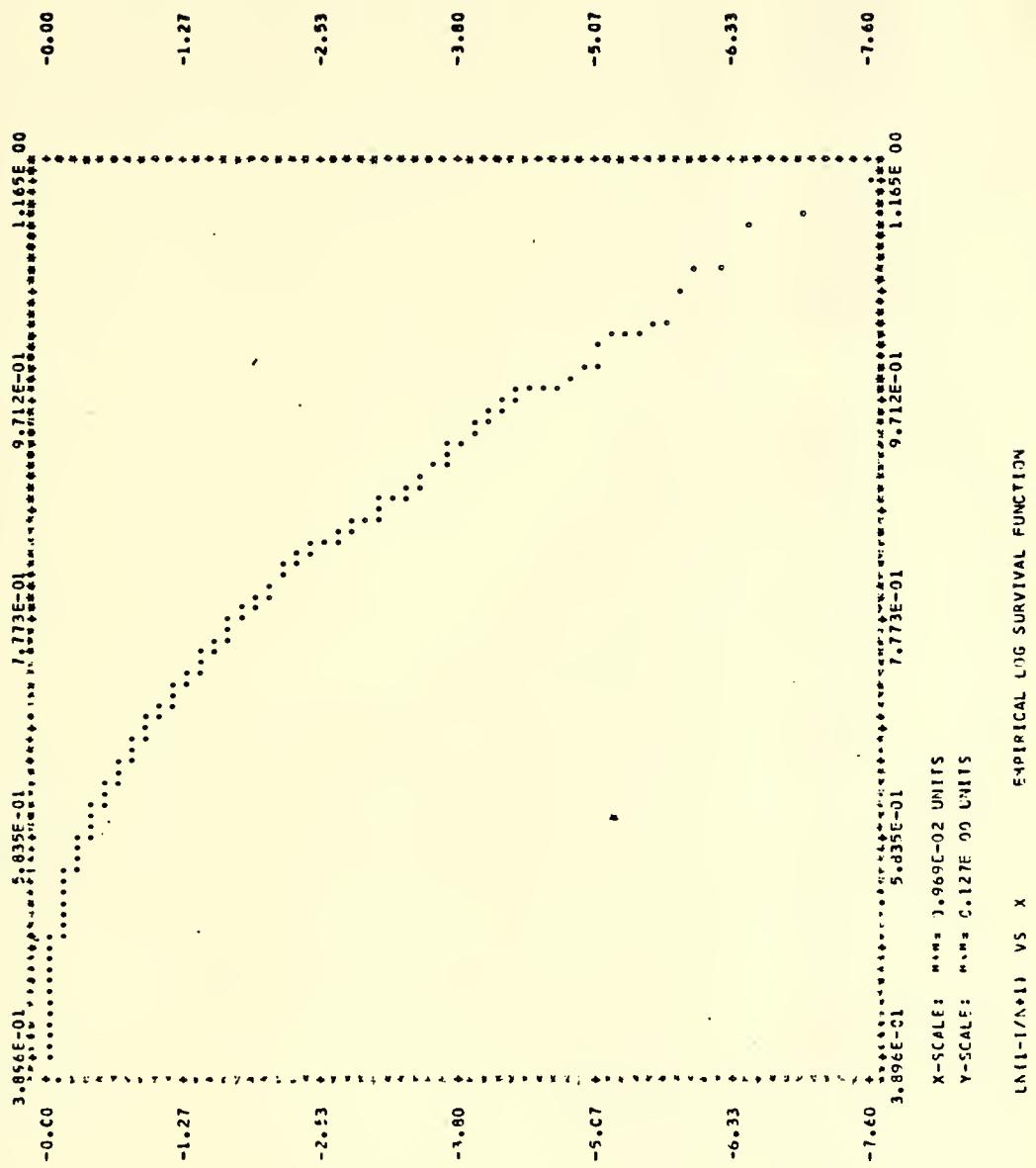
INTERVAL SQUARE NORM SAMPLE SIZE:  $N = 2000$   
 UNIF 100 RANDOM VARIABLE SAMPLE SIZE  $N = 1500$   
 TRIANGULAR  $\approx$  INCH. BANDWIDTH =  $1/\lambda$



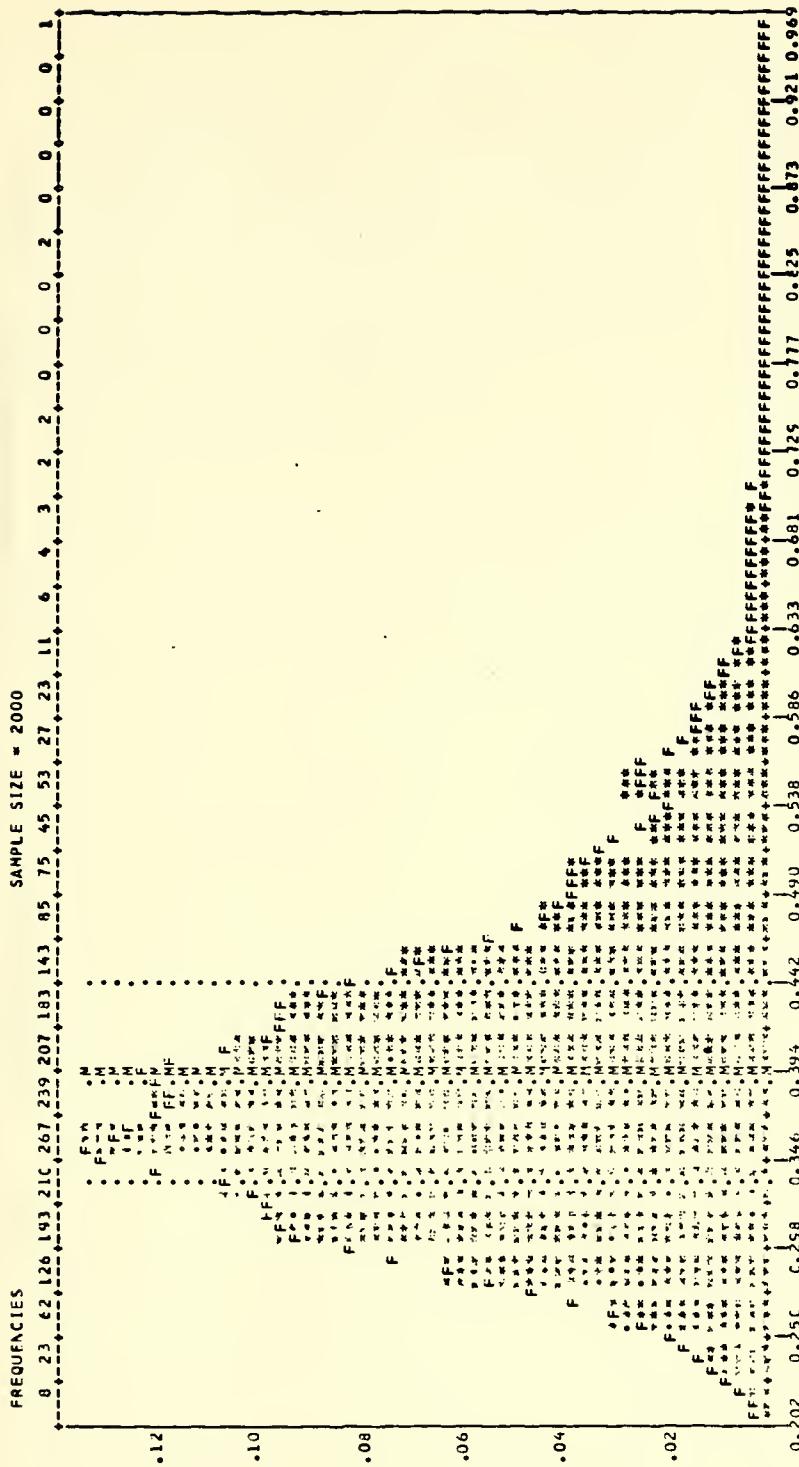


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS





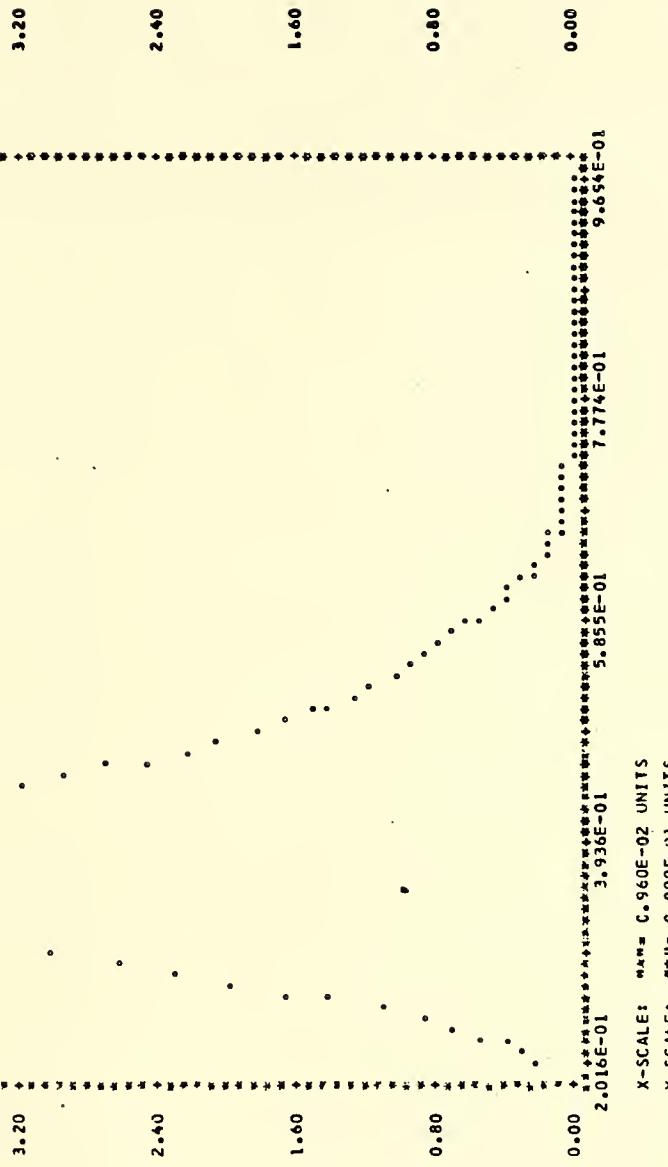




CENTRAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS				DISTRIBUTION
		VARIANCE	H <sub>3</sub>	H <sub>4</sub>	M <sub>3</sub>	
MEAN	3.920465E-01	7.12676E-03	5.94738E-04	6.94474E-04	2.01670E-0	2.015897E-0
MEAN	3.920465E-01	8.52165E-02	6.51198E-01	6.51198E-01	2.02343E-0	2.02343E-0
MEAN	3.920465E-01	2.05794E-01	2.05794E-01	2.05794E-01	2.02343E-0	2.02343E-0
MEAN	3.920465E-01	7.05712E-01	7.05712E-01	7.05712E-01	2.02343E-0	2.02343E-0
MEAN	3.920465E-01	1.05952E-01	1.05952E-01	1.05952E-01	2.02343E-0	2.02343E-0
MEAN	3.920465E-01	2.25884E-01	2.25884E-01	2.25884E-01	2.02343E-0	2.02343E-0
MEAN	3.920465E-01	2.25884E-01	2.25884E-01	2.25884E-01	2.02343E-0	2.02343E-0

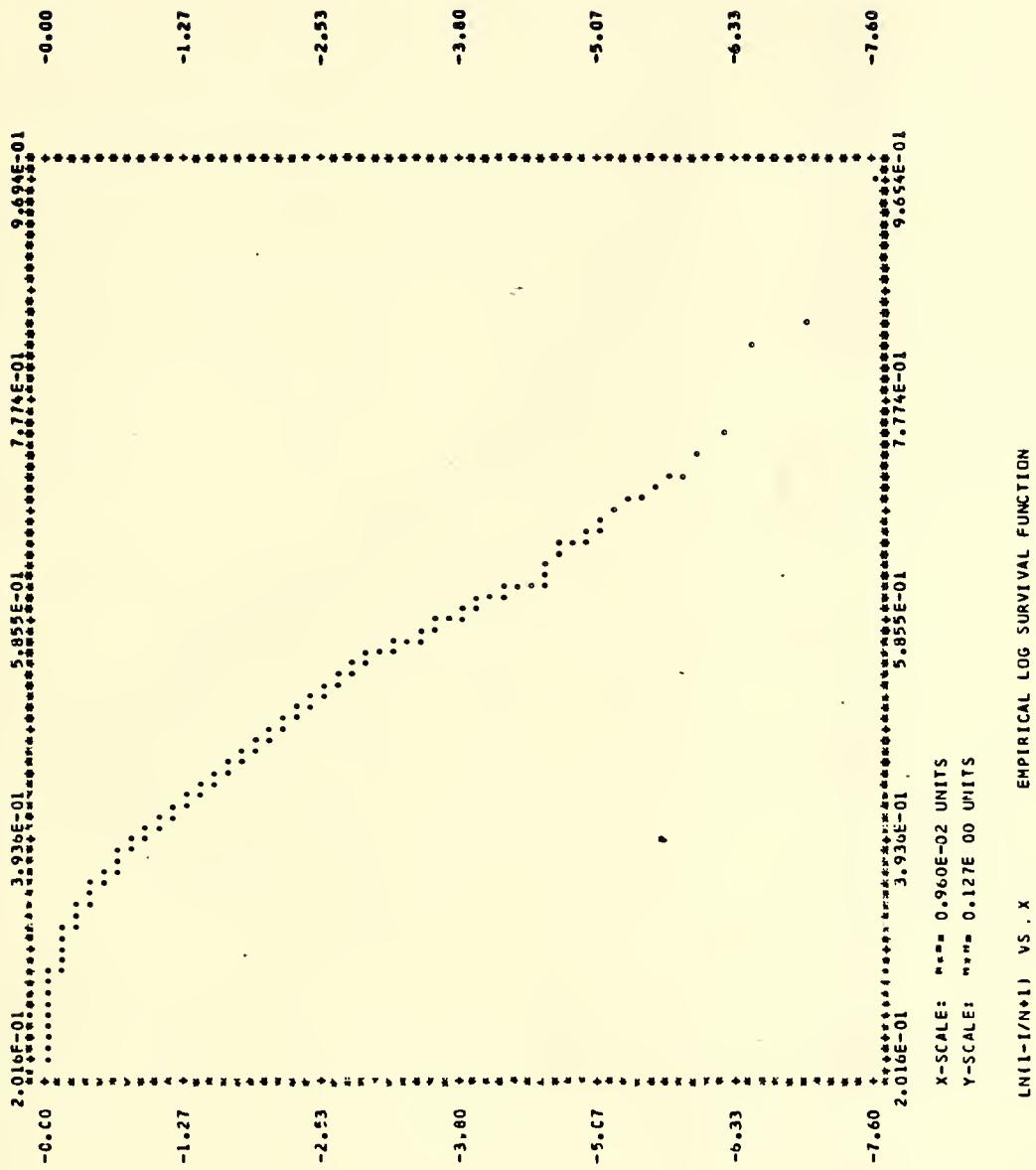
INTEGRAL SQUARE NORM SAMPLE SIZE  $n = 2000$   
 CAUCHY RANDOM VARIABLE SAMPLE SIZE  $n = 100$   
 TRIANGULAR WINDOW. EACH WIDTH =  $1/SQRT(n)$



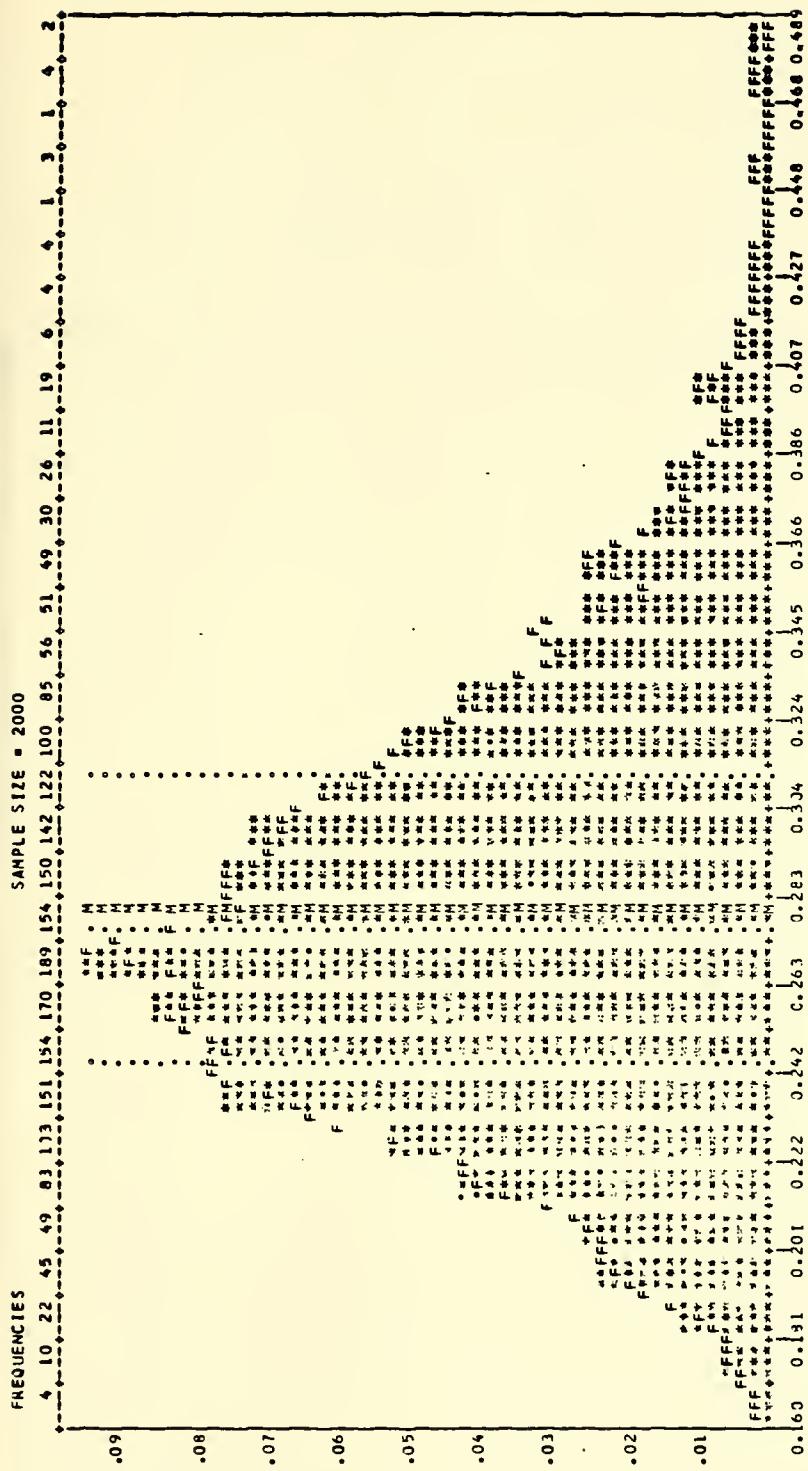


ESTIMATED CENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED CENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS







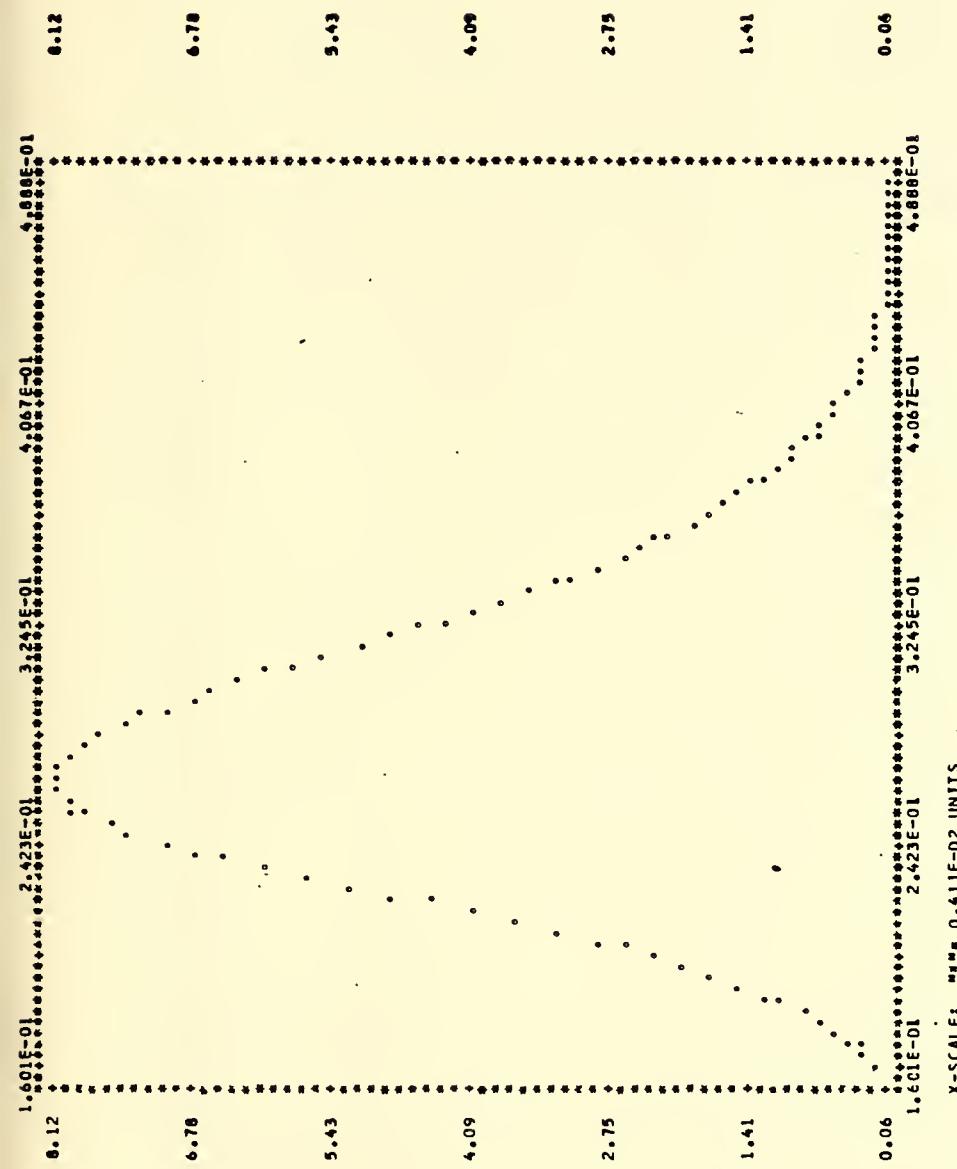


#### CENTRAL TENDENCY

	SPREAD	HIGHER CENTRAL MOMENTS	DISTRIBUTION
MEAN	2.800428E-01	M3 7.3225448E-05	MINIMUM 1.601461E-01
MEDIAN	2.718137E-01	M4 2.225913E-05	10 QUANTILE 1.20366E-01
TRIMMED MEAN	2.715929E-01	SKENNESS 5.894413E-01	25 QUANTILE (HINGE) 2.44097E-01
MEAN	2.159345E-01	KURTOSIS 5.924845E-01	50 QUANTILE (MEDIAN) 2.74037E-01
GEOM. MEAN	2.15226E-01		75 QUANTILE (HINGE) 3.10361E-01
HARM. MEAN	2.15021E-01		90 QUANTILE 3.45594E-01
			MAXIMUM 4.88842E-01

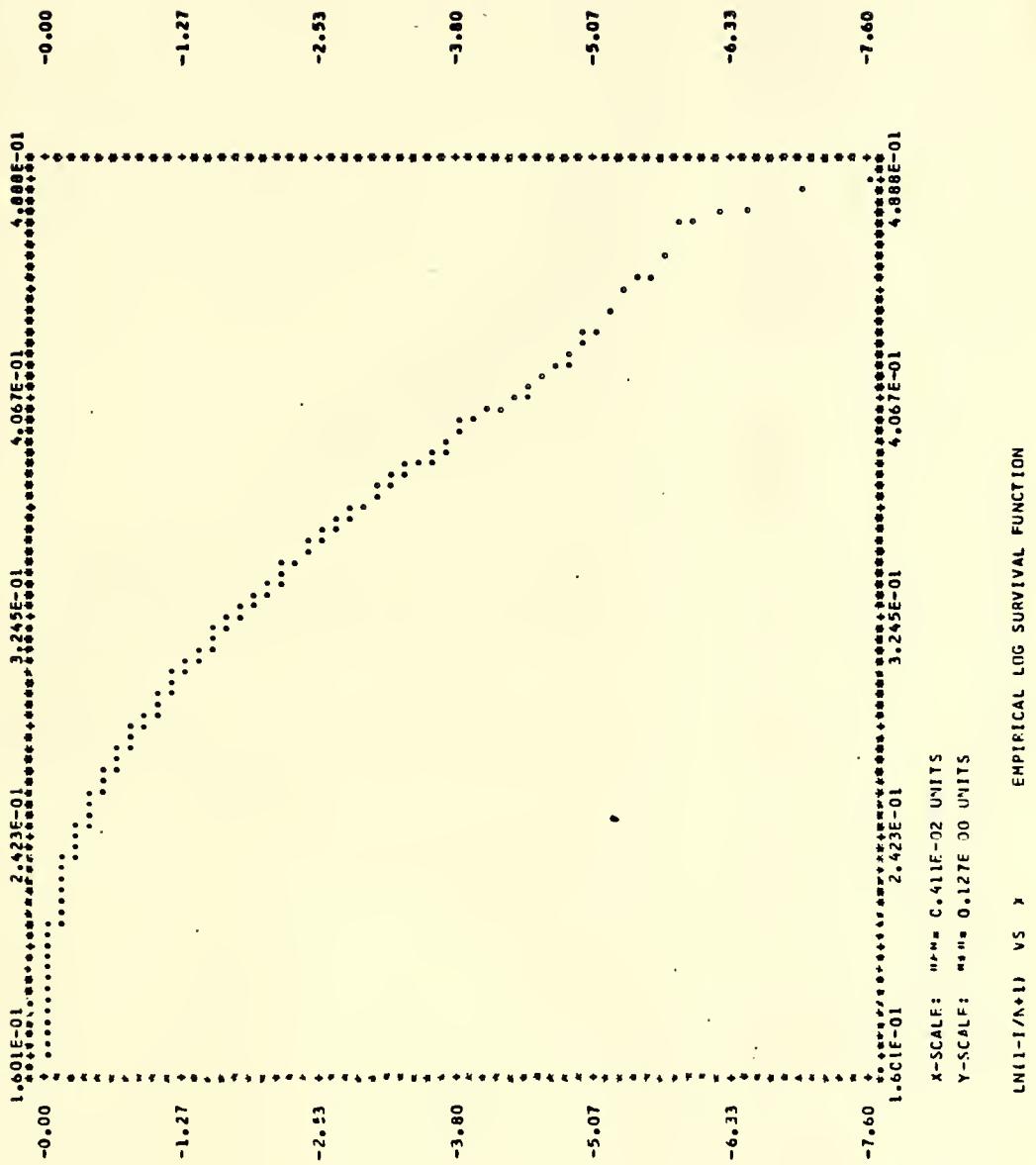
INTEGRAL SQUARE 'NORM' SAMPLE SIZE M = 2000  
 CAUCHY RANDOM VARIABLE SAMPLE SIZE N = 200  
 TRIANGULAR WINDOW. BANDWIDTH = 1/SQRT(N)





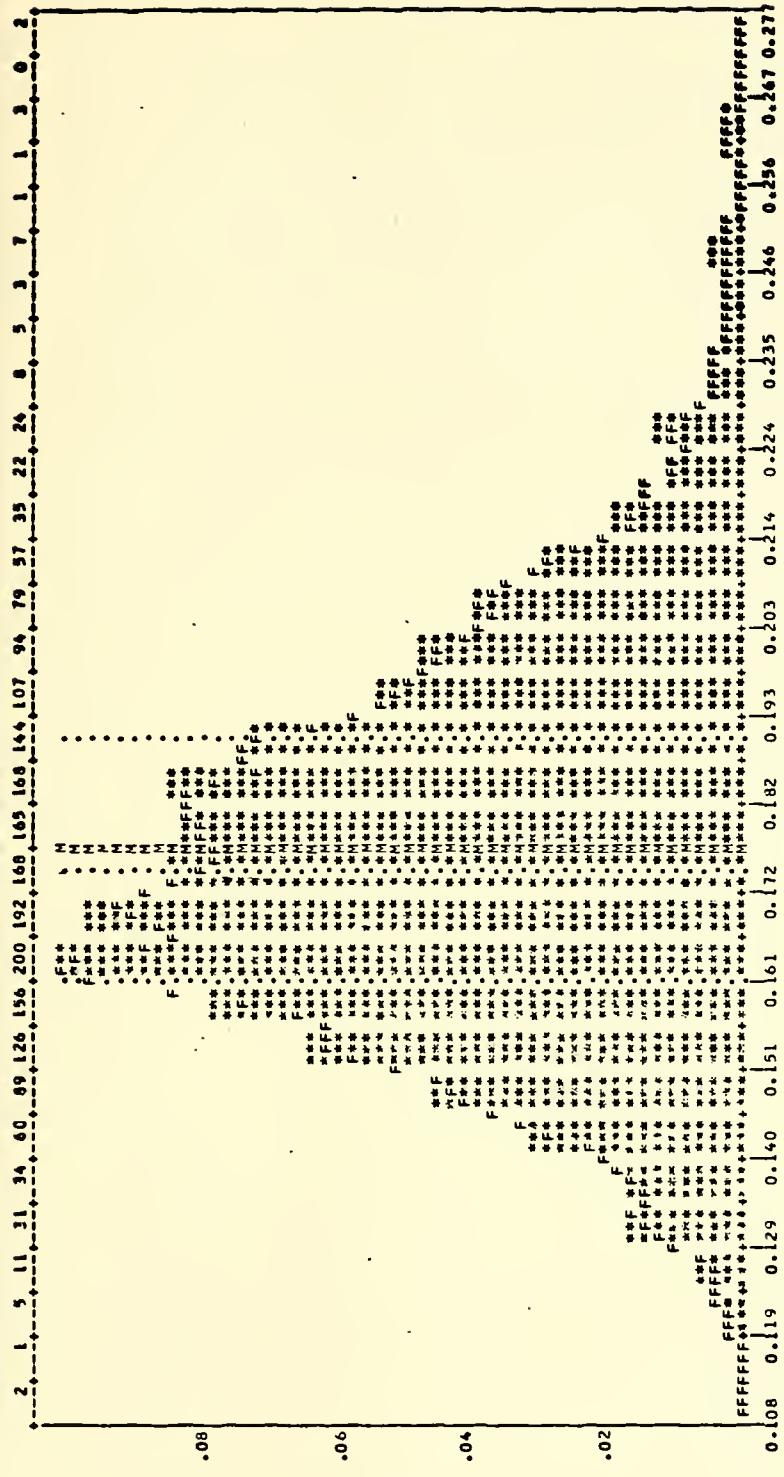
ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED CENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS







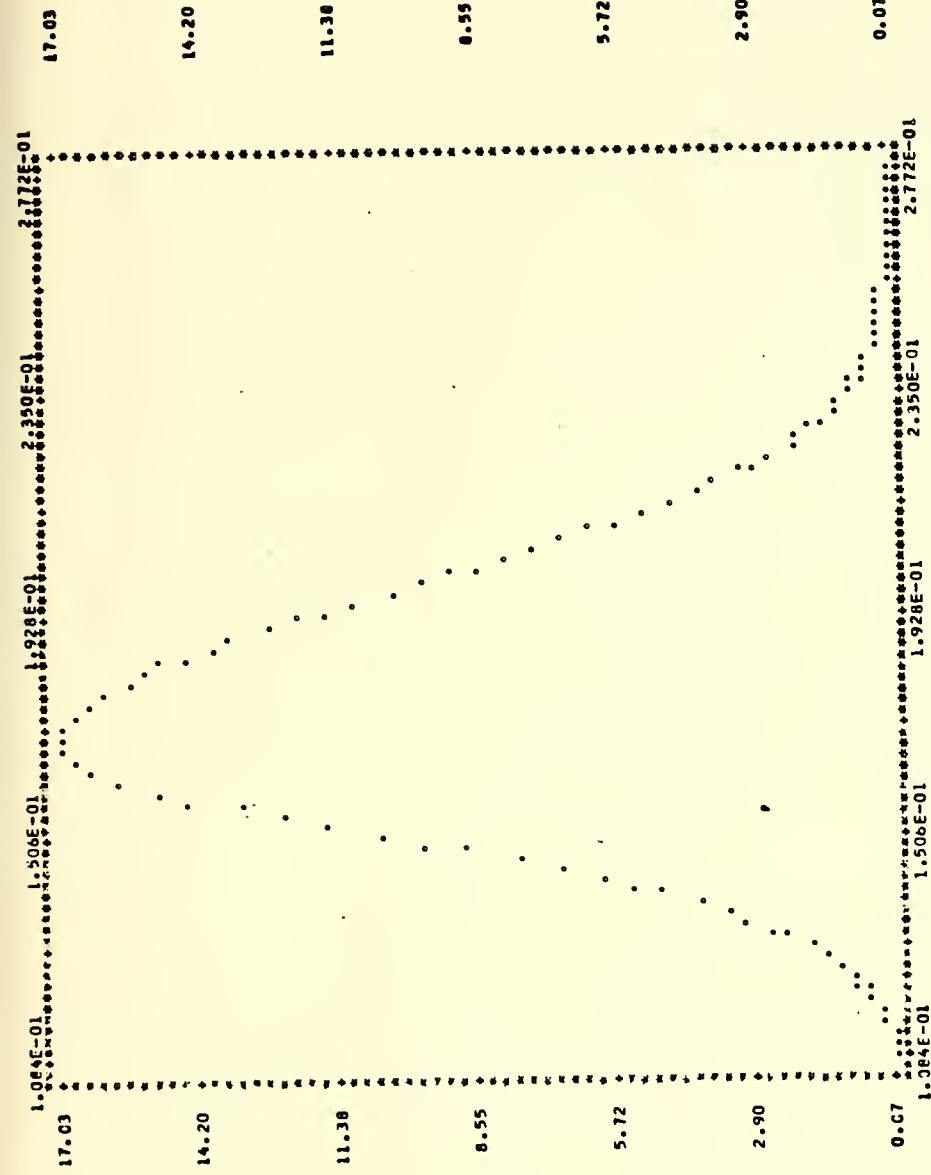
SAMPLE SIZE = 2000



CENTRAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS	DISTRIBUTION
MEAN 1.763979E-01	VARIANCE 5.243721E-04	M3 5.346300E-06	MINIMUM 0.839231E-01
MEDIAN 1.716314E-01	STD DEV 2.289971E-02	M4 9.660602E-07	10 QUANTILE 4.87713E-01
TRIMMED 1.751256E-01	COEF VAR 1.298554E-01	SKENNESS 4.450774E-01	25 QUANTILE 6.07134E-01
MILOWIAN 1.755445E-01	RANGE 1.6881722E-01	KURTOSIS 5.093794E-01	50 QUANTILE 7.63445E-01
GECH MEAN 1.749359E-01	SPAN 2.981722E-02		75 QUANTILE 9.02856E-01
HARM MEAN 1.734888E-01			90 QUANTILE 2.06078E-01
			MAXIMUM 2.77235E-01

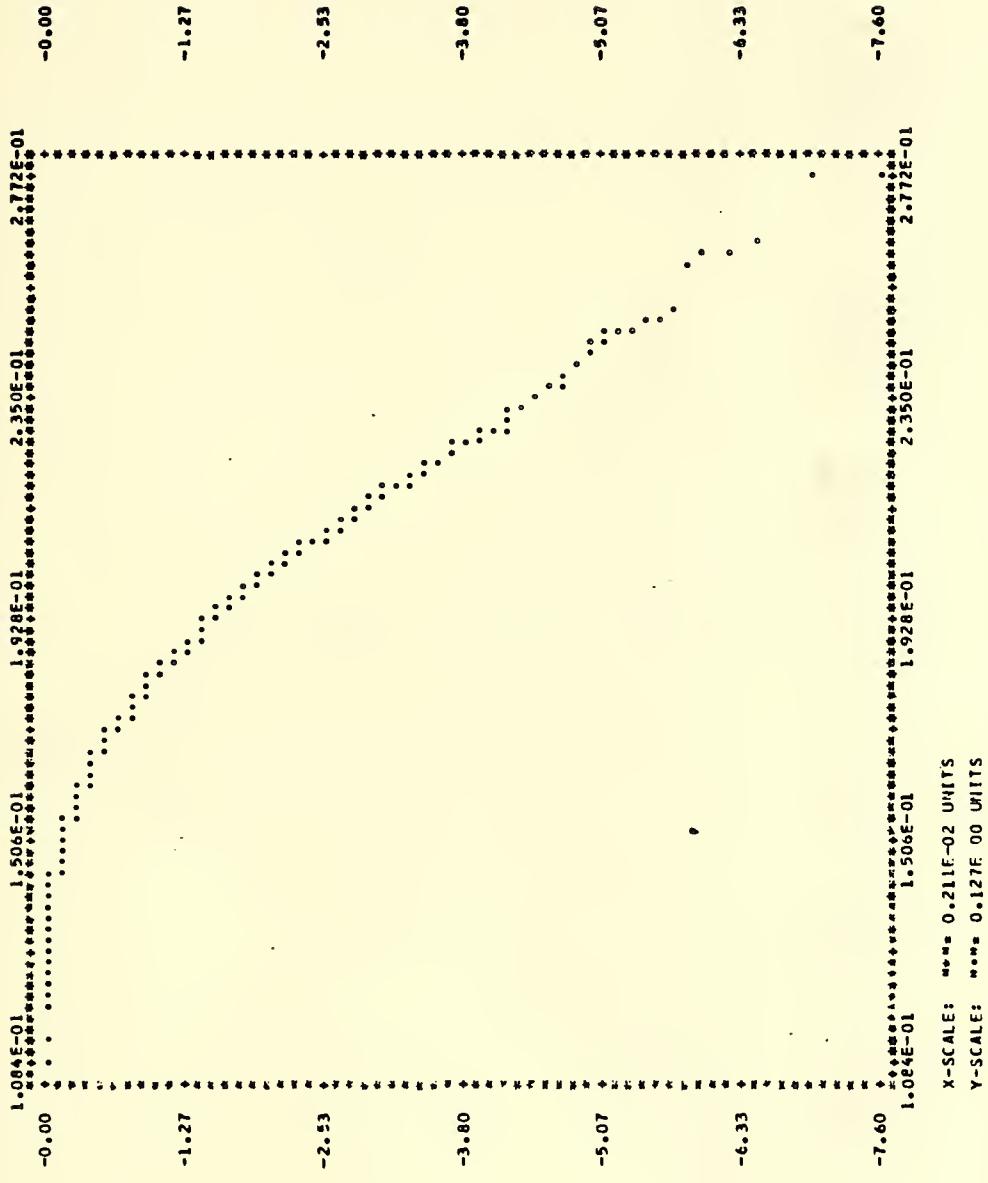
INTEGRAL SQUARE NORM SAMPLE SIZE M = 2000  
CAUCHY RANDOM VARIABLE SAMPLE SIZE N = 500  
TRIANGULAR WINDOW. BANDWIDTH = 1/SQRT(N)





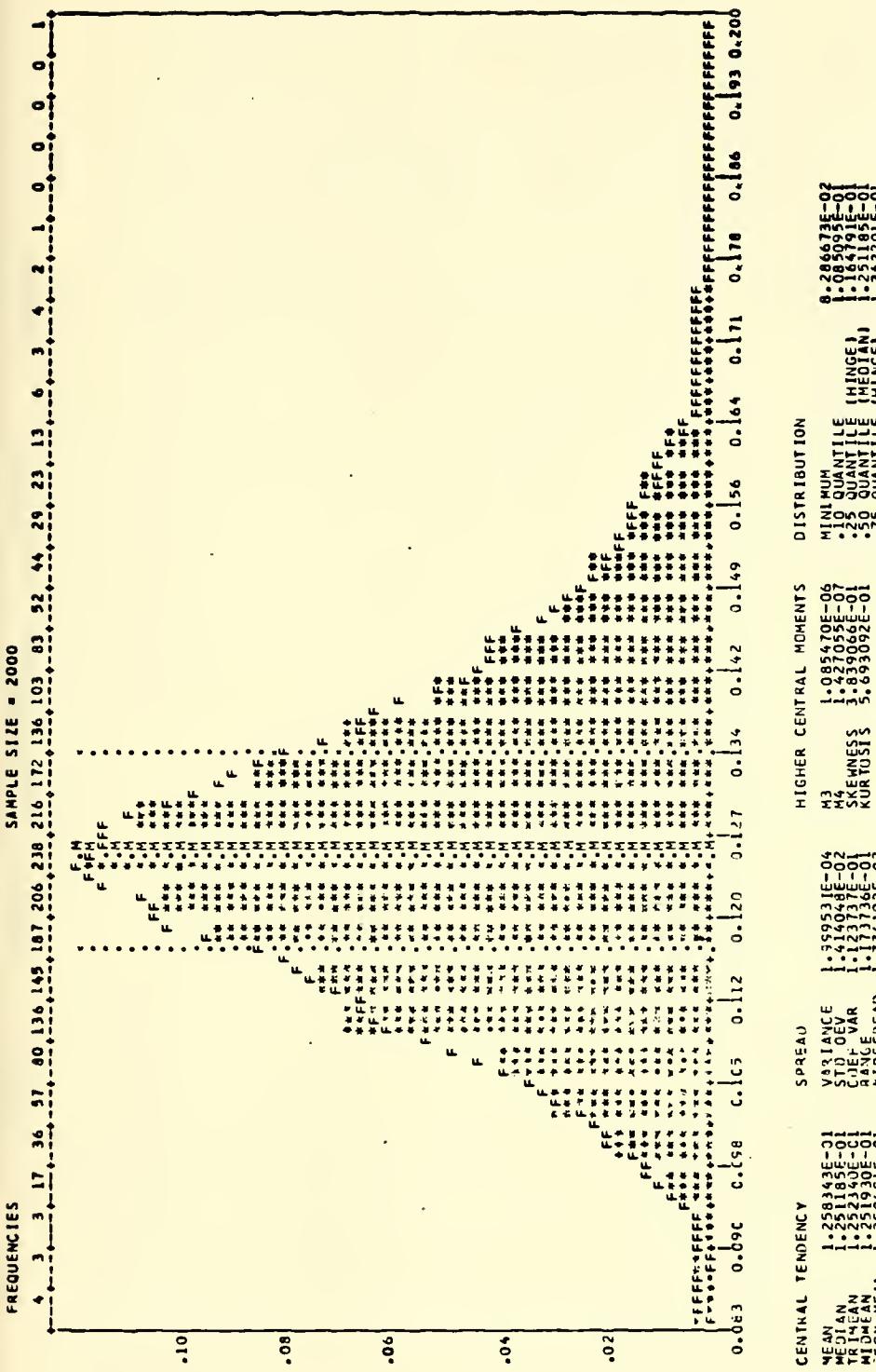
ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000      TRIANGULAR WINDOW      BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS





LN(1/(N+1)) VS X      EMPIRICAL LOG SURVIVAL FUNCTION





INTEGRAL SQUARE NCPM SAMPLE SIZE  $\eta = 2000$   
 CAUCHY P-ANOM VARIABLE SAMPLE SIZE  $N = 1000$   
 TRIANGULAR BOUND FANOUCI $\chi^2$  VSORTINI

CENTRAL TENDENCY

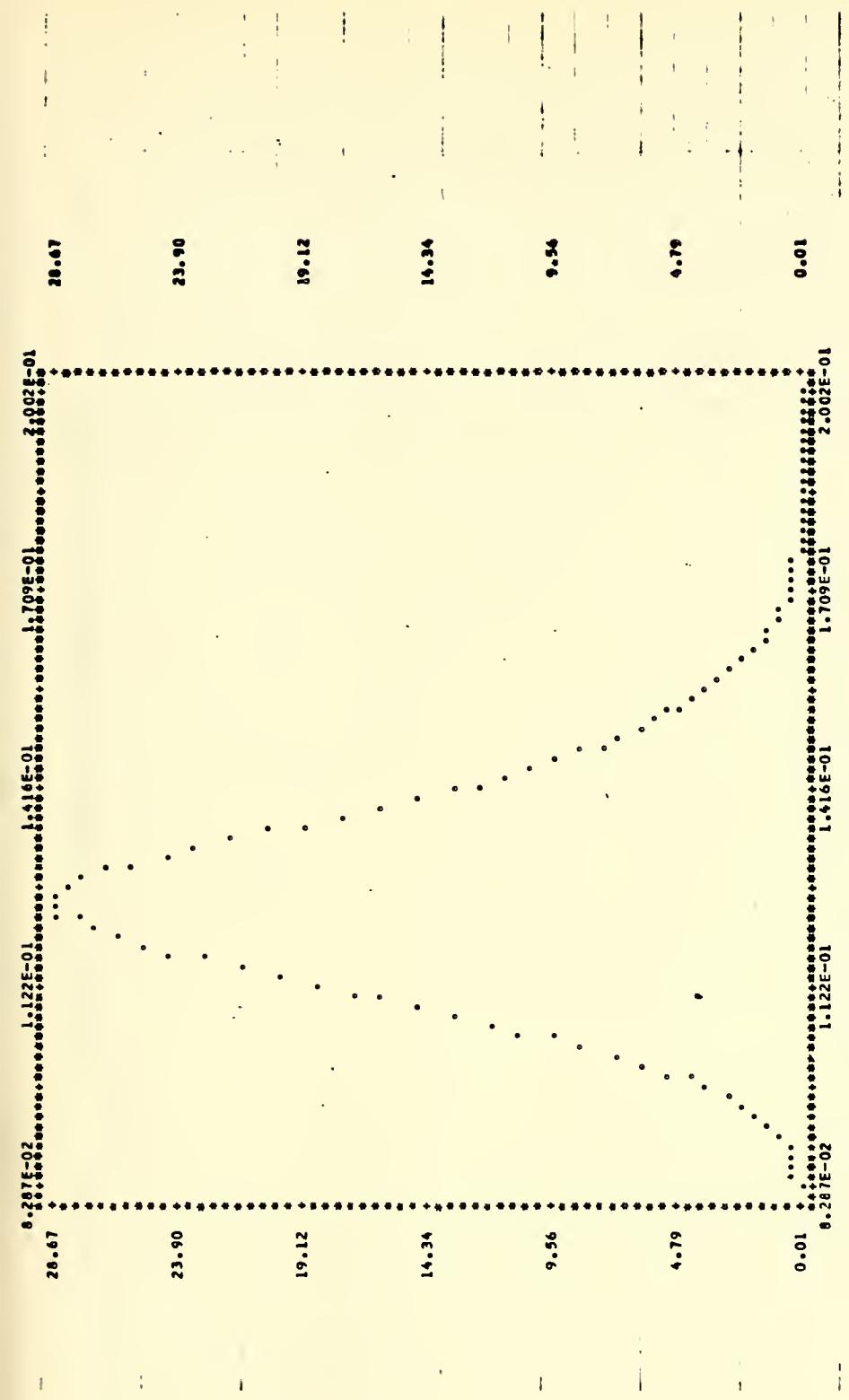
	SPEA U	V AR IANCE	STD OVE R	C OEFF VAR	R AACE FER	PIC FER EAD	HIGHER CENTRAL MOMENTS
		$3.95 \times 10^{-4}$	$1.02$	$-0.01$	$-0.01$	$-0.01$	$M_3 = 1.0844705 \times 10^0$
		$1.49 \times 10^{-4}$	$1.27$	$-0.01$	$-0.01$	$-0.01$	$M_4 = 1.0420550 \times 10^0$
		$1.74 \times 10^{-4}$	$1.74$	$-0.01$	$-0.01$	$-0.01$	$\text{SKEWNESS} = 5.6933092 \times 10^{-1}$
							$\text{KURTOSIS} = 5.6933092 \times 10^{-1}$

HIGHER CENTRAL MOMENTS DISTRIBUTION

SAMPLE SIZE = 2000

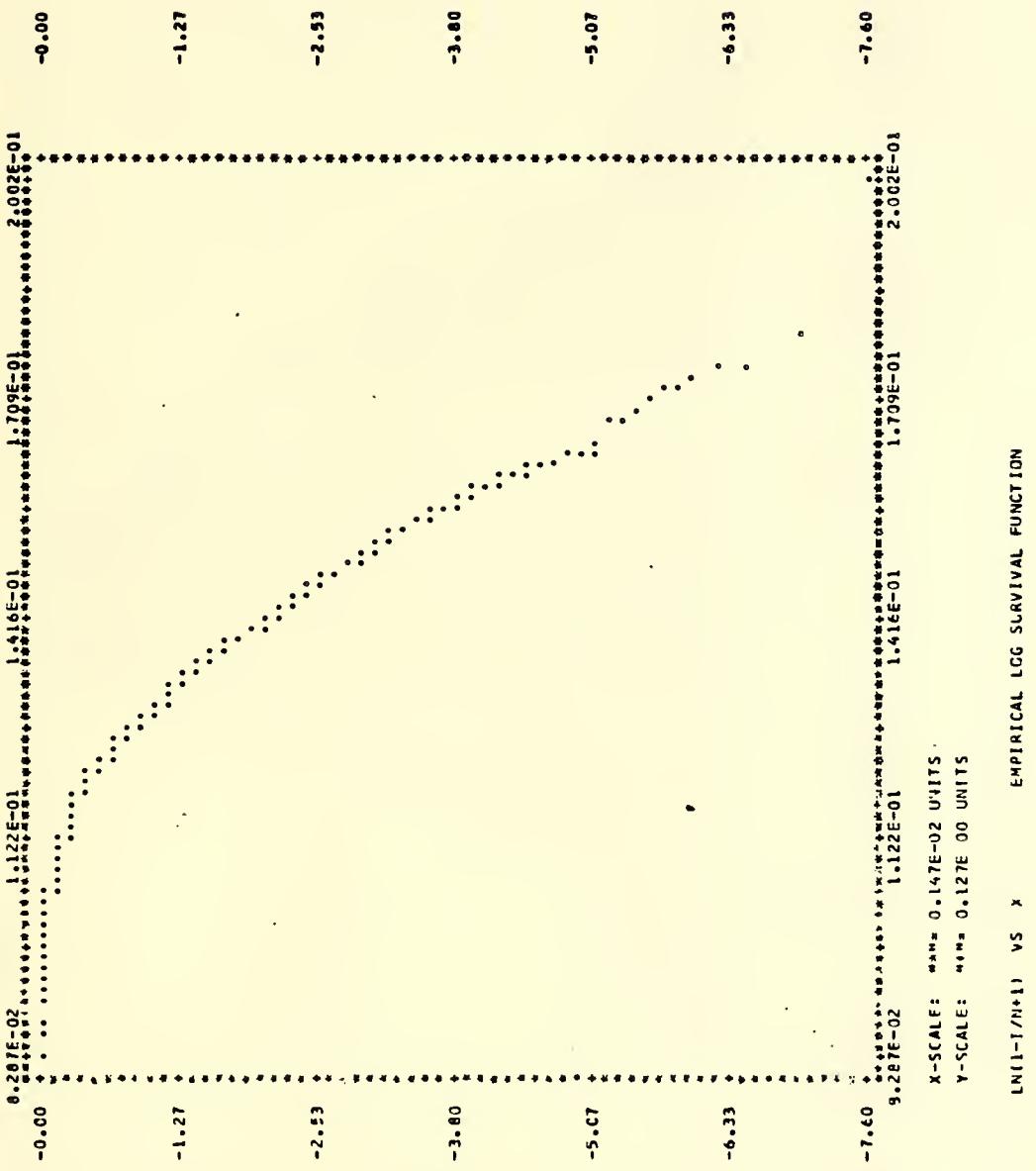
FREQUENCIES



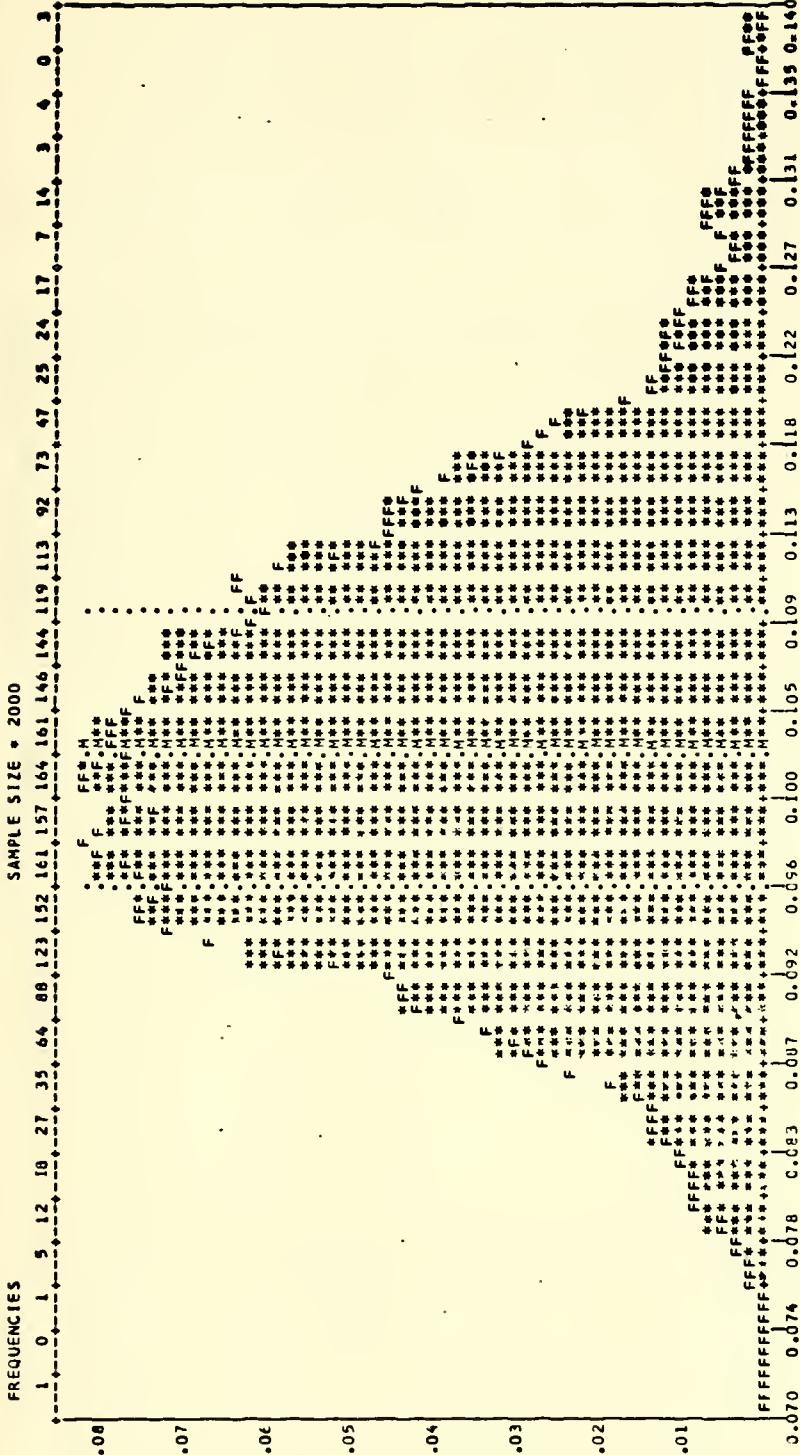


ESTIMATED CENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000    TRIANGULAR WINDOW    RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS







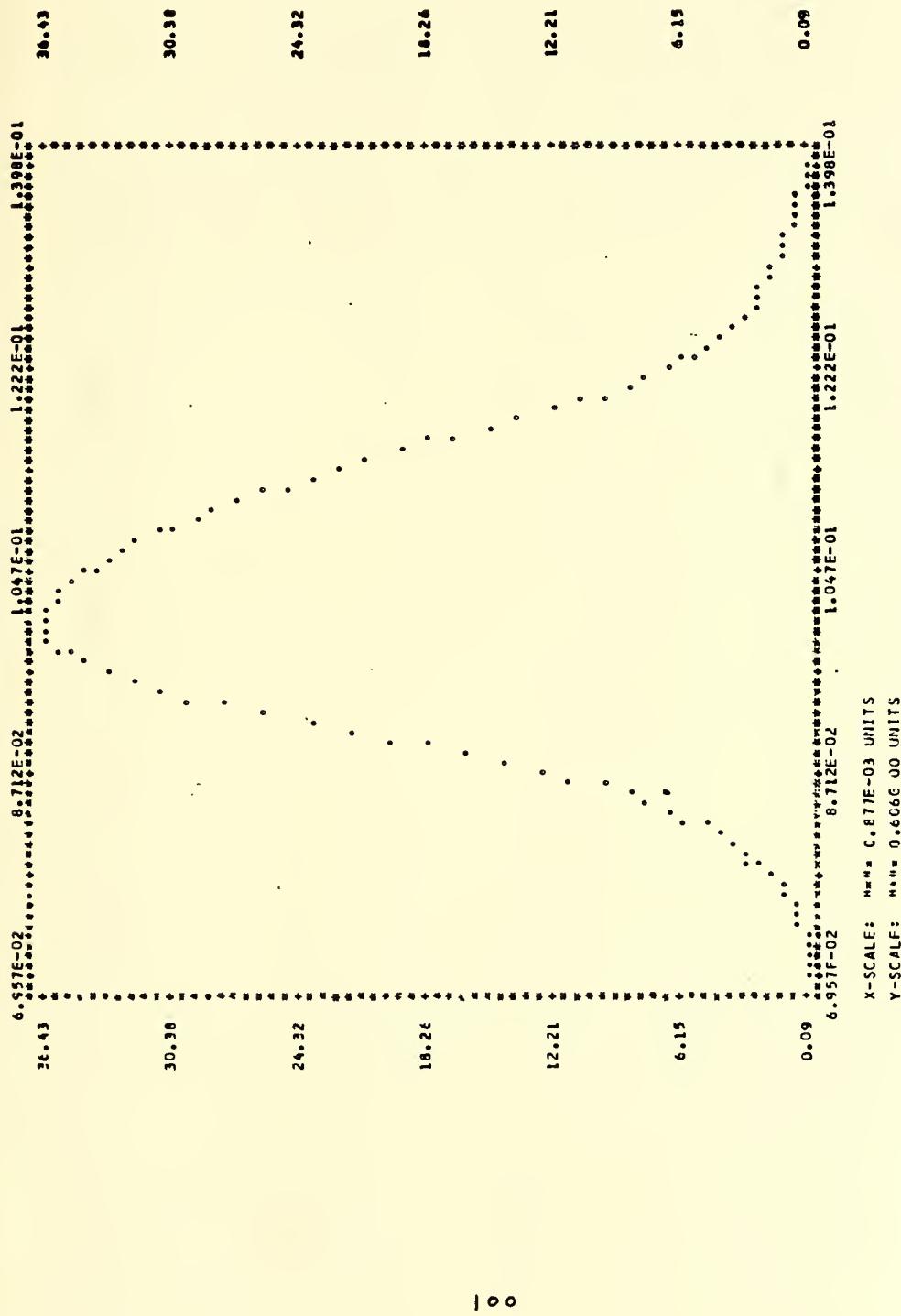


HIGHER CENTRAL MOMENTS

CENTRAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS
MEAN	1.028634E-01	M3 1.049400E-04
MEDIAN	1.021599E-01	M4 2.663205E-07
TRIMMEAN	1.021503E-01	10 QUANTILE 3.324776E-08
MIDMEAN	1.02144C8E-01	25 QUANTILE 2.477281E-01
GGM MEAN	1.021360E-01	KURTOSIS 1.857281E-02
HARM MEAN	1.018693E-01	75 QUANTILE 1.098802E-01
		90 QUANTILE 1.16063E-01
		MAXIMUM 1.39766E-01

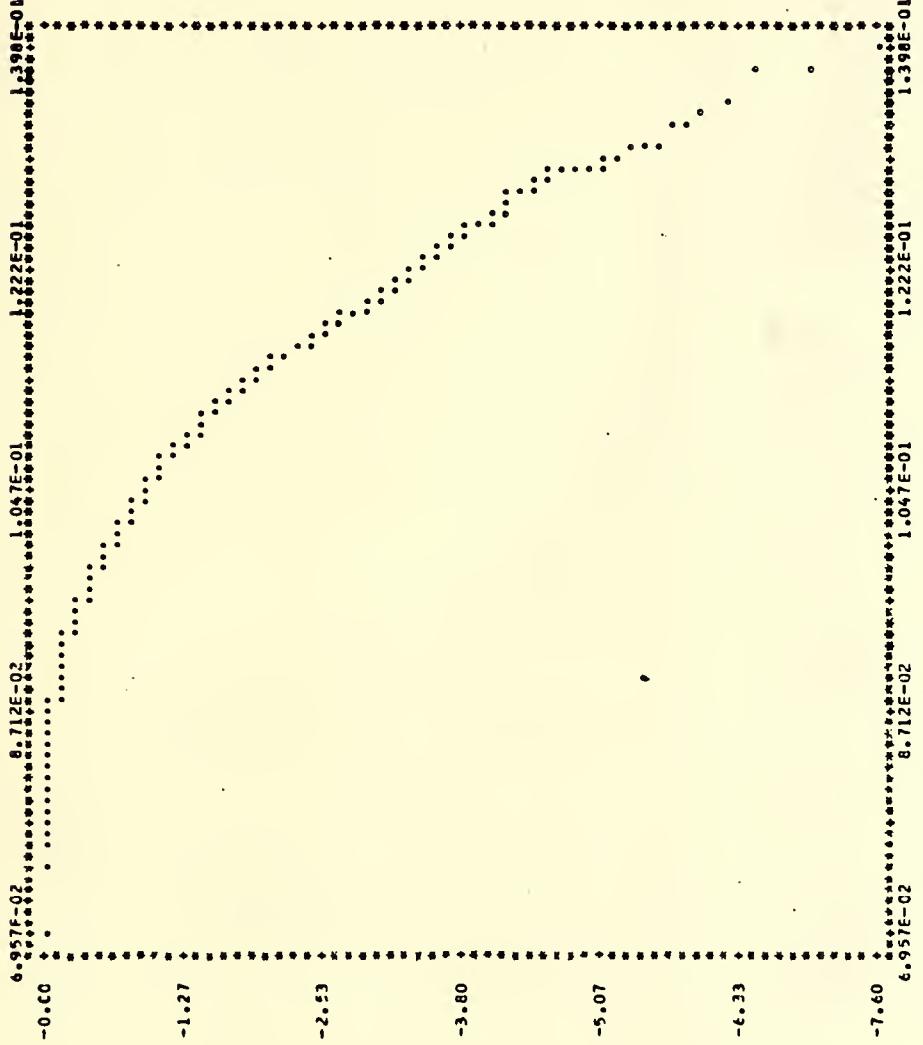
INTEGRAL SQUARE NORM SAMPLE SIZE N = 2000  
 CAUCHY RAND VAR VARIABLE SAMPLE SIZE N = 1500  
 TRIANGULAR WINDOW. BANDWIDTH = 1/SQRT(N)





ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000      TRIANGULAR WINDOW      BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS

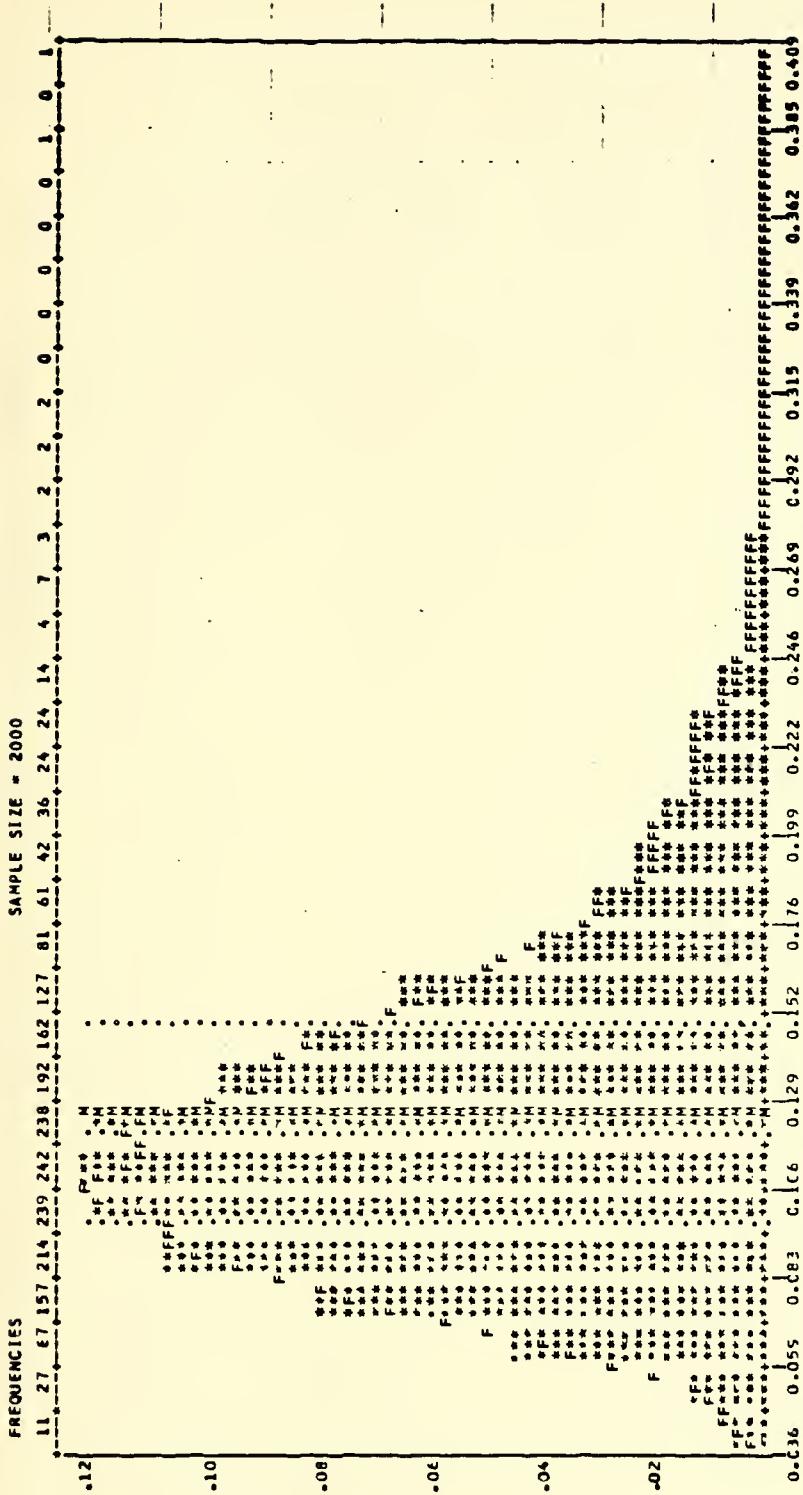




2

LN(1 - 1/N+1) VS X EMPIRICAL LOG SURVIVAL FUNCTION

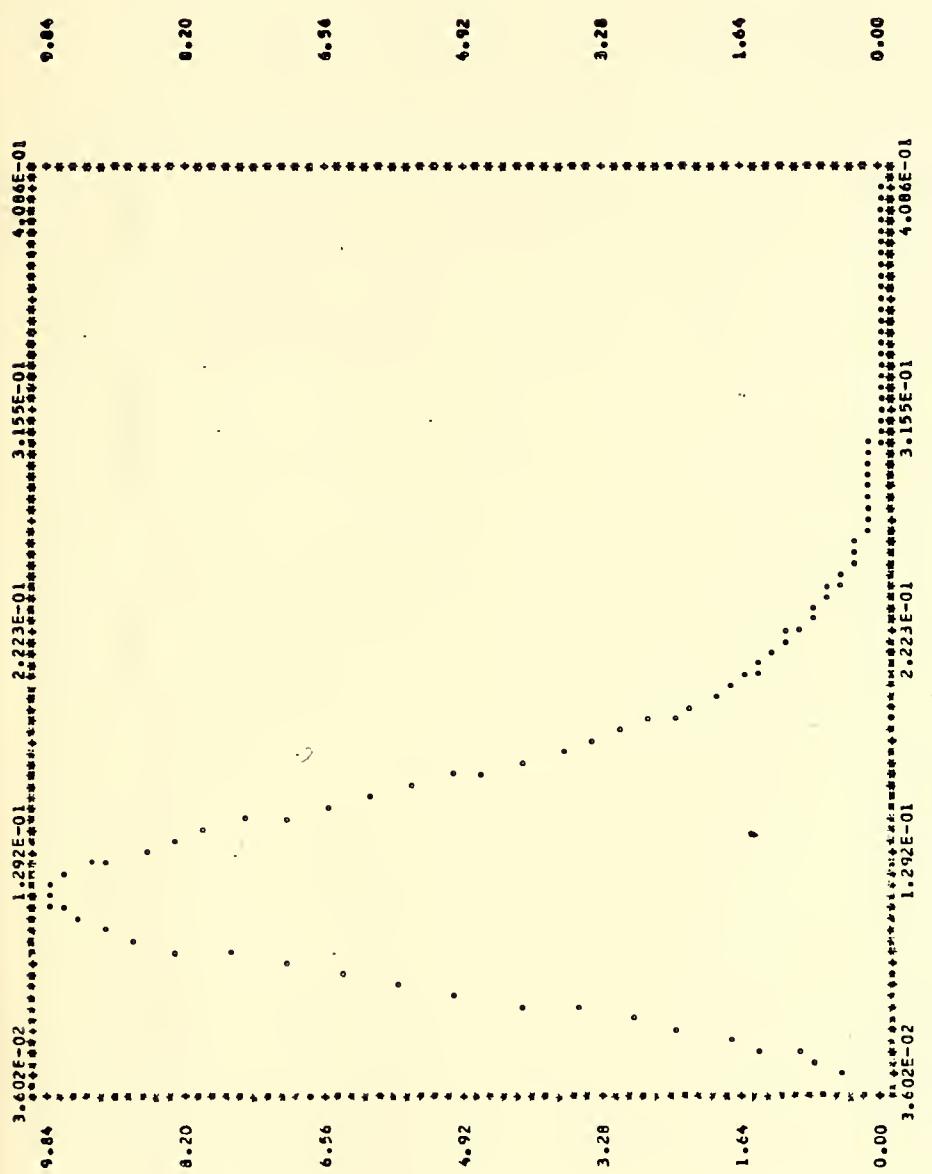




CENTRAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS			DISTRIBUTION
		VARIANCE	STD DEV	SKEWNESS	
MEAN	1.247724E-01	1.804262E-03	M3	8.461093E-05	MINIMUM
MEAN	1.878CE-C	4.404262E-02	M4	1.7450E-05	0 QUANTILE
MEAN	1.8941E-01	3.405331E-01	COEF VAR	1.04020E-05	25 QUANTILE
MEAN	1.9296E-01	3.72665E-01	STDEV	2.49078E-05	50 QUANTILE
MEAN	1.8C71E-01	5.31632E-02	KURTOSIS	2.49078E-05	75 QUANTILE
MEAN	1.117418E-01	1.0515E-01	SKWNESS	0.0	90 QUANTILE
MEAN	1.117418E-01	1.0515E-01	KURTOSIS	0.0	MAXIMUM

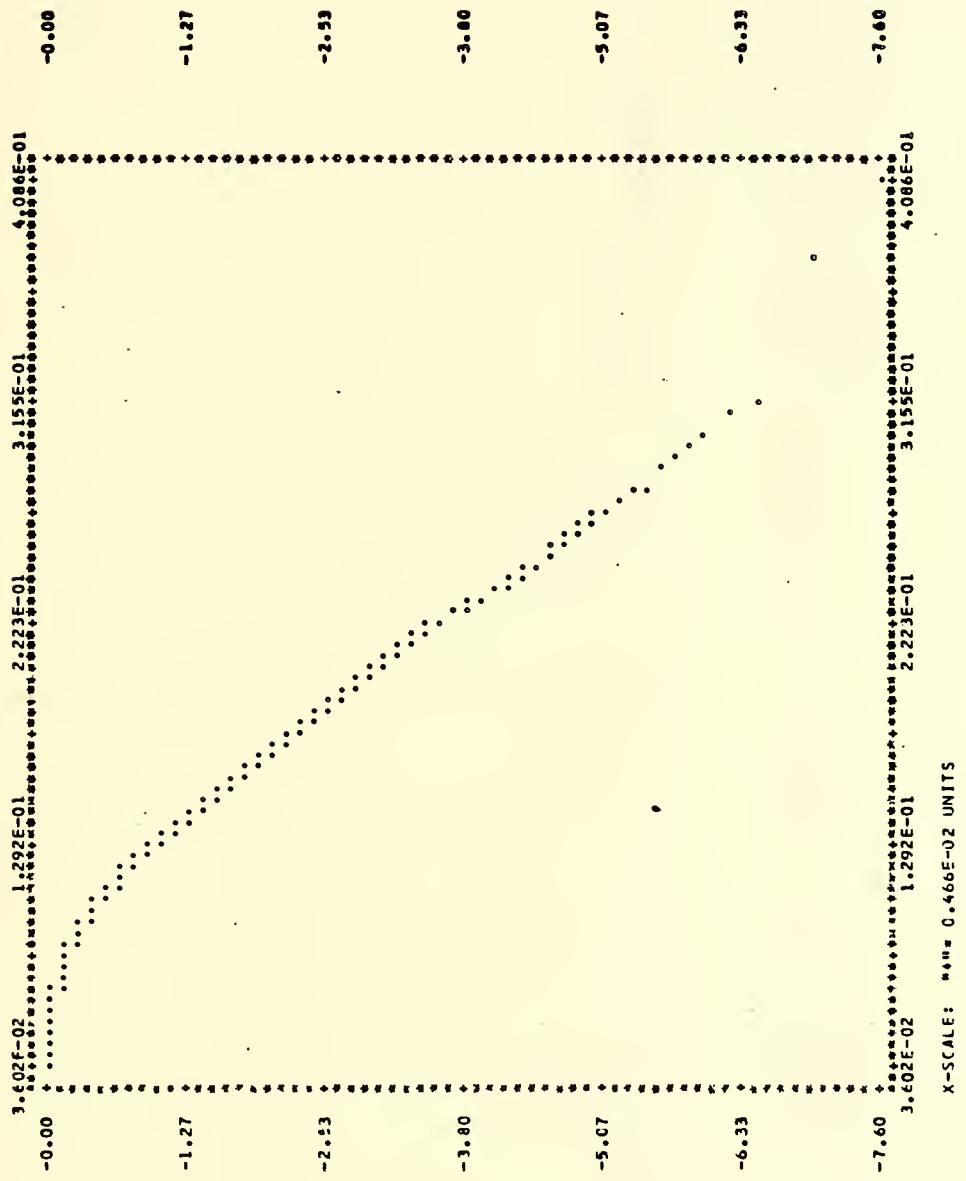
INTEGRAL SCLARE HIGH SAMPLE SIZE P = 2000  
 CAUCHY PANOCP VARIABLE SAMPLE SIZE N = 100  
 TRIANGULAR BIASCH EANHICHT = 3/SQRT(N)





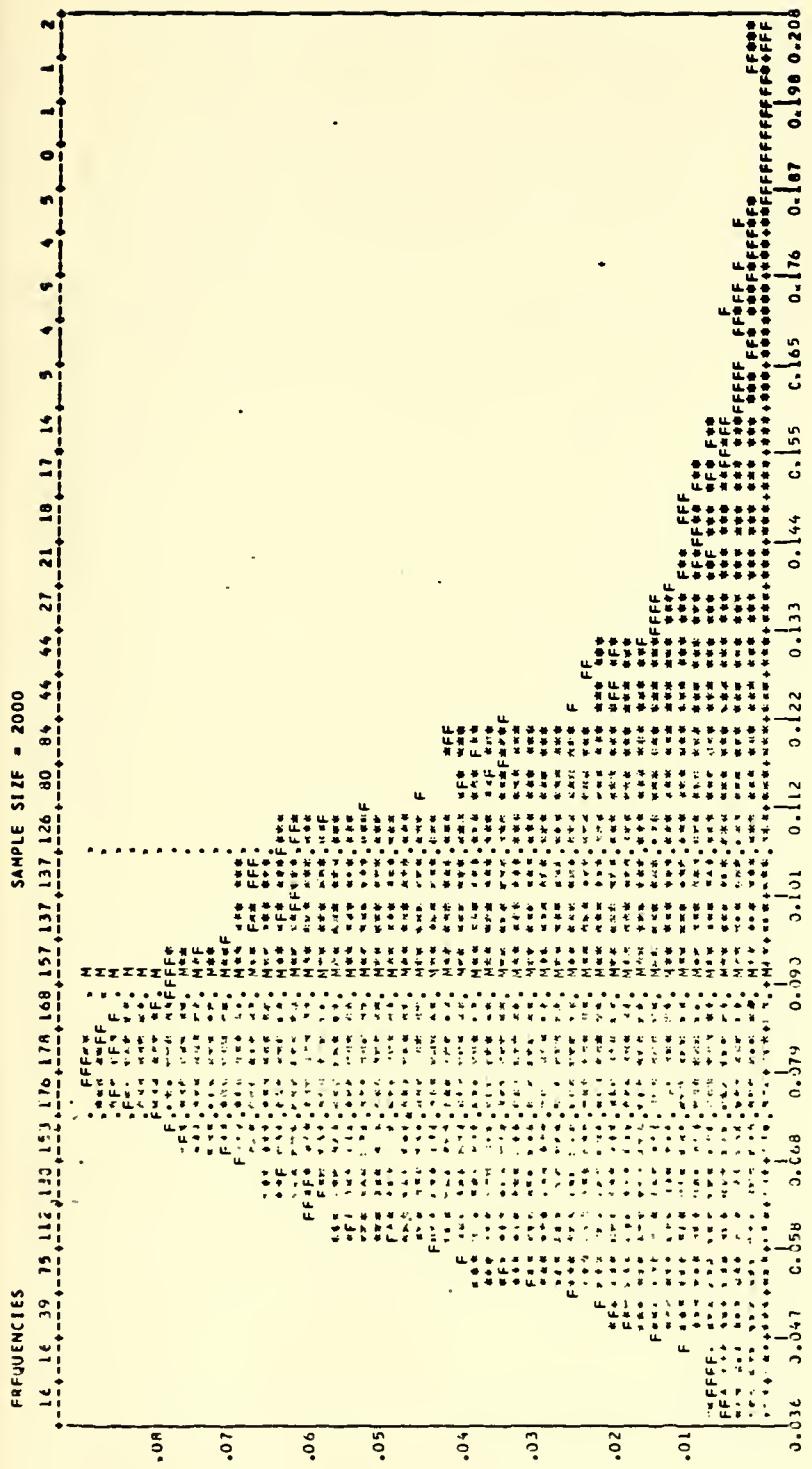
ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS





10<sup>-4</sup>



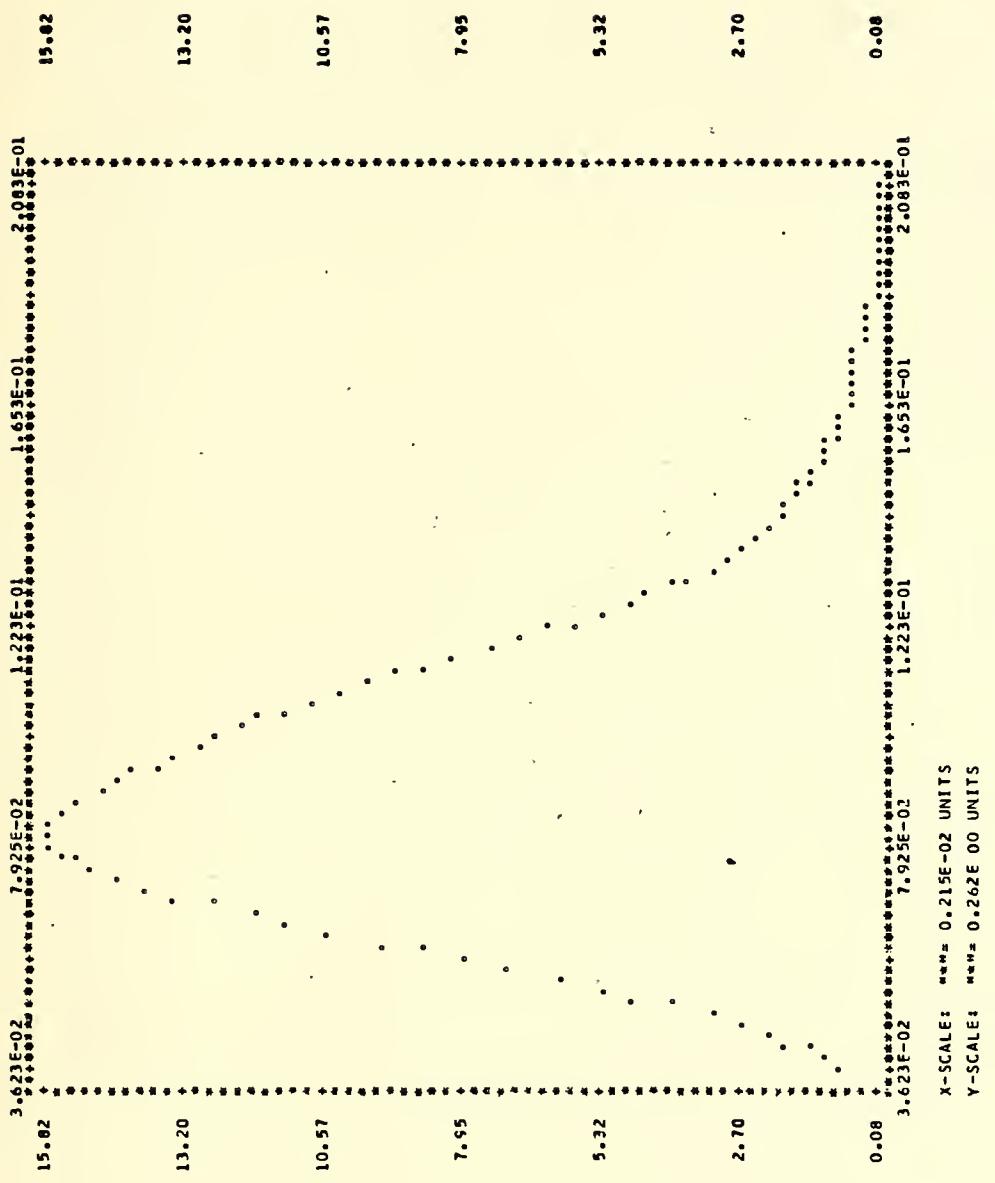


INTEGRAL SQUARE NORM SAMPLE SIZE $N = 2000$	CENTERAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS	DISTRIBUTION
MEAN	$9.19521E-02$	VARIANCE	$6.72957E-04$	$M_3 = 1.354832E-05$
MEAN	$8.03450E-02$	STD DEV	$2.096579E-02$	MINIMUM
TRIANGULAR	$8.82534E-02$	COVARIANCE	$2.096175E-01$	25 QUANTILE (THINNESS)
TRIANGULAR	$8.82534E-02$	KURTOSIS	$1.743546E-01$	50 QUANTILE (THICKNESS)
GAUSSIAN	$8.74259E-02$	MOOD	$3.351714E-01$	90 QUANTILE (THICKNESS)
HARMONIC	$8.405149E-02$	PERCENTILE	$0.165$	HARITHM

INTEGRAL SQUARE NORM SAMPLE SIZE  $N = 2000$   
 CAUCHY RANDOM VARIABLE SAMPLE SIZE  $N = 200$   
 TRIANGULAR INNOV.

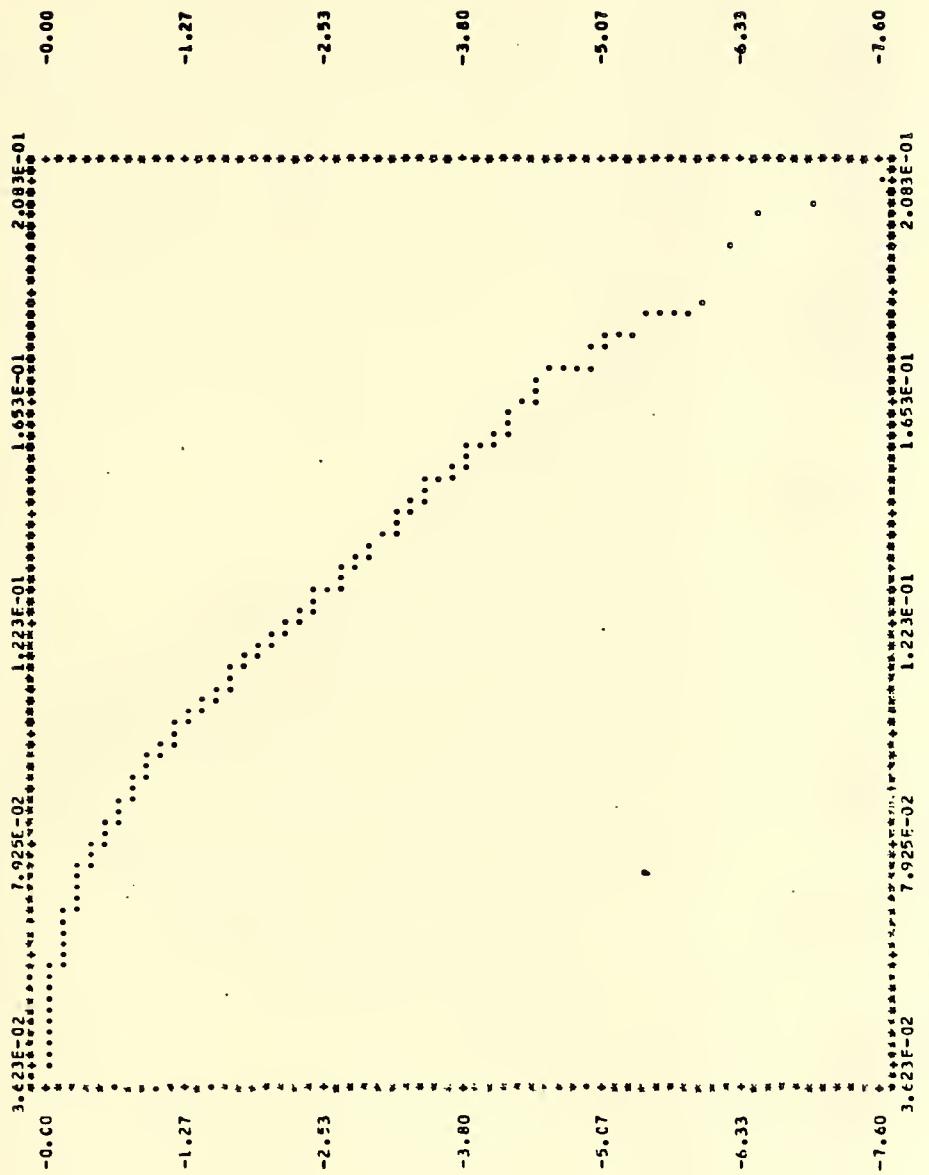
PENULTIMATE =  $3/\sqrt{N}$



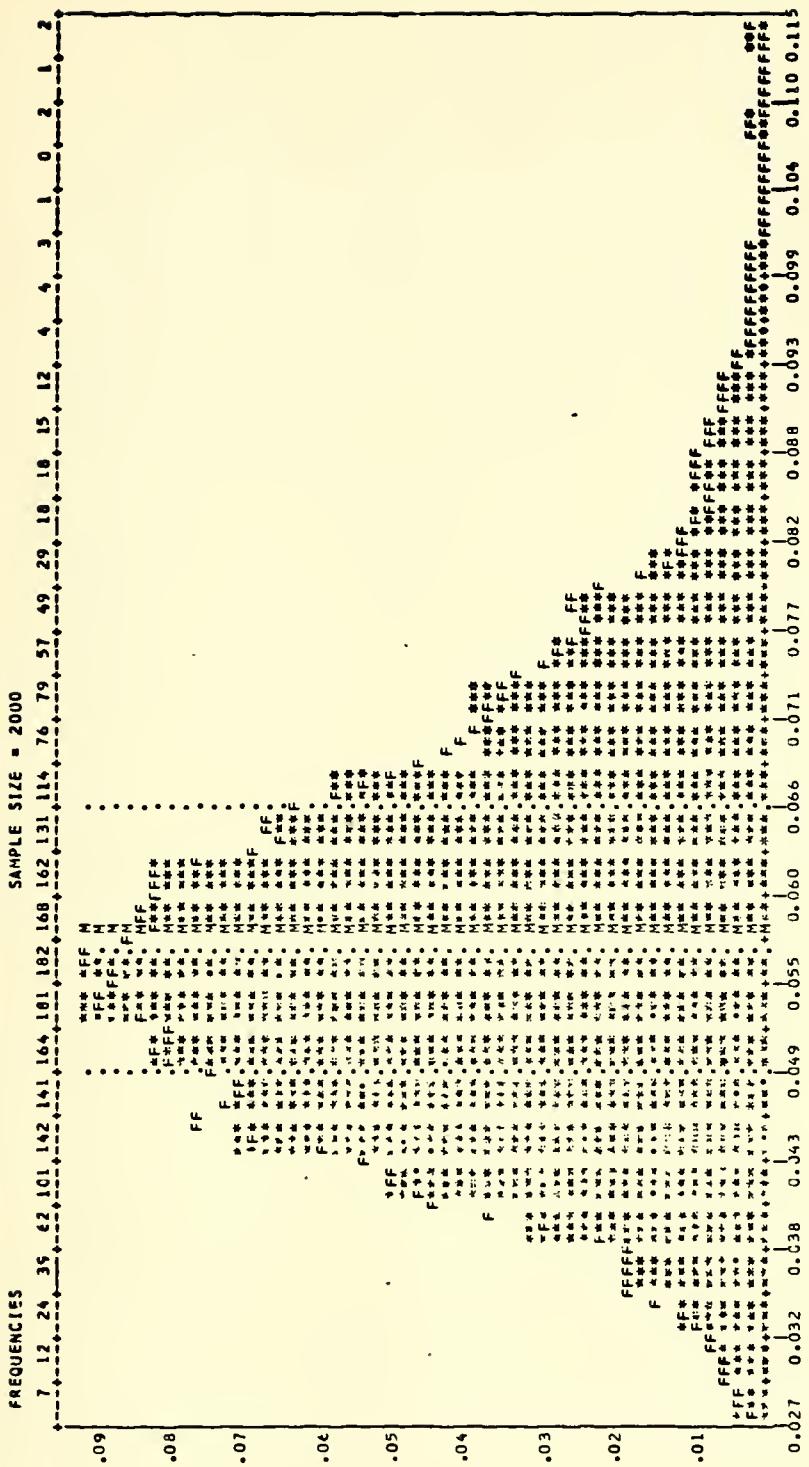


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS





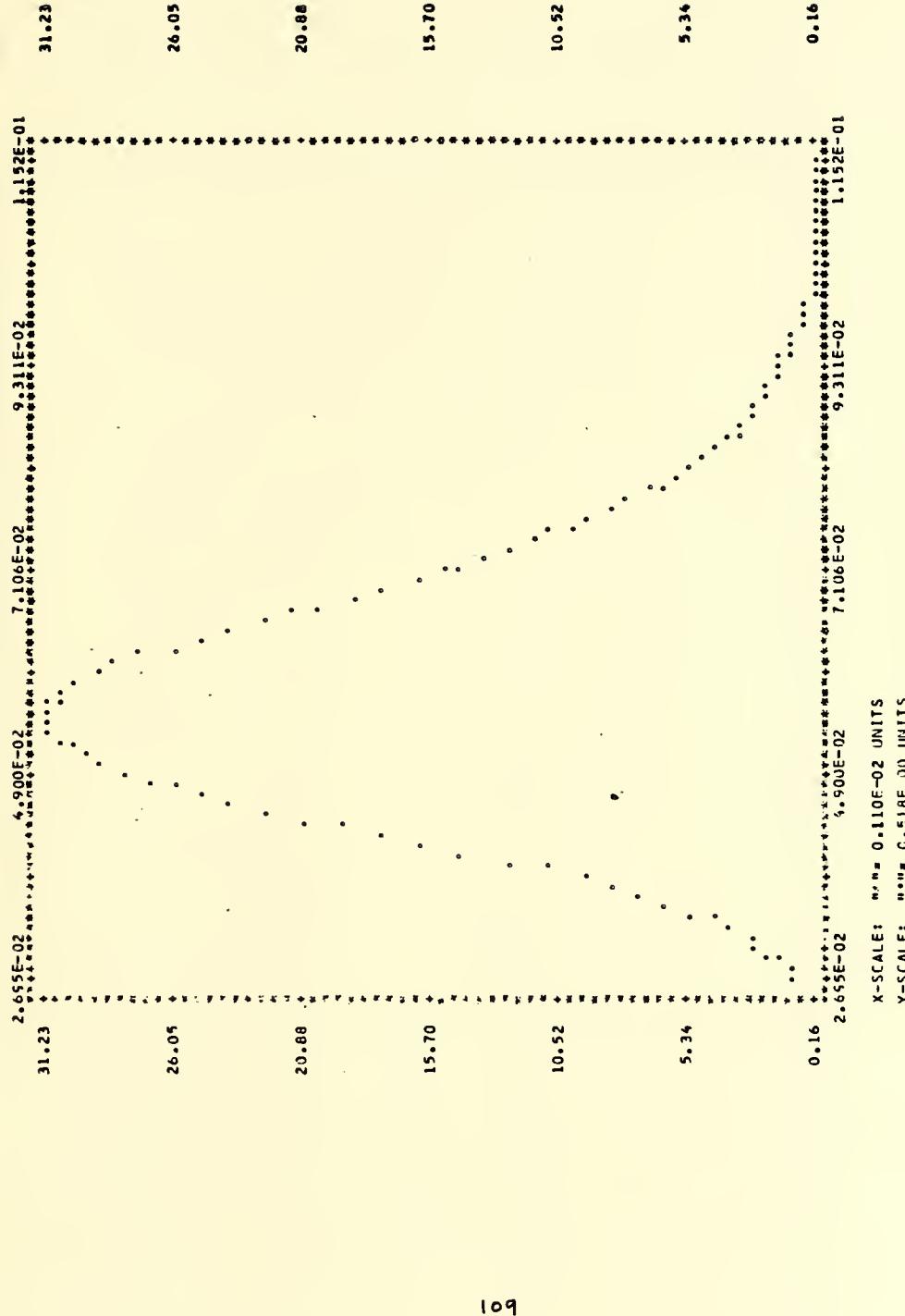




CENTRAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS			DISTRIBUTION
		M3	M4	KURTOSIS	
MEAN	5.762135E-02	1.676320E-02	1.24404E-02	1.0241E-02	MINIMUM
MEDIAN	5.449246E-02	1.278831E-02	9.95428E-03	8.1021E-03	QUANTILE
TRIMMED MEAN	5.664141E-02	1.310931E-02	1.05448E-02	8.5026E-03	(HNGE)
HODGEAN MEAN	5.66927195E-02	1.310931E-02	1.05448E-02	8.5026E-03	(MDIAN)
GEOM. MEAN	5.623759E-02	1.26525E-02	9.8566E-03	7.75Q	QUANTILE
HARM. MEAN	5.4867759E-02	1.26525E-02	9.8566E-03	1.151682E-01	(HNGE)
					MAXIMUM

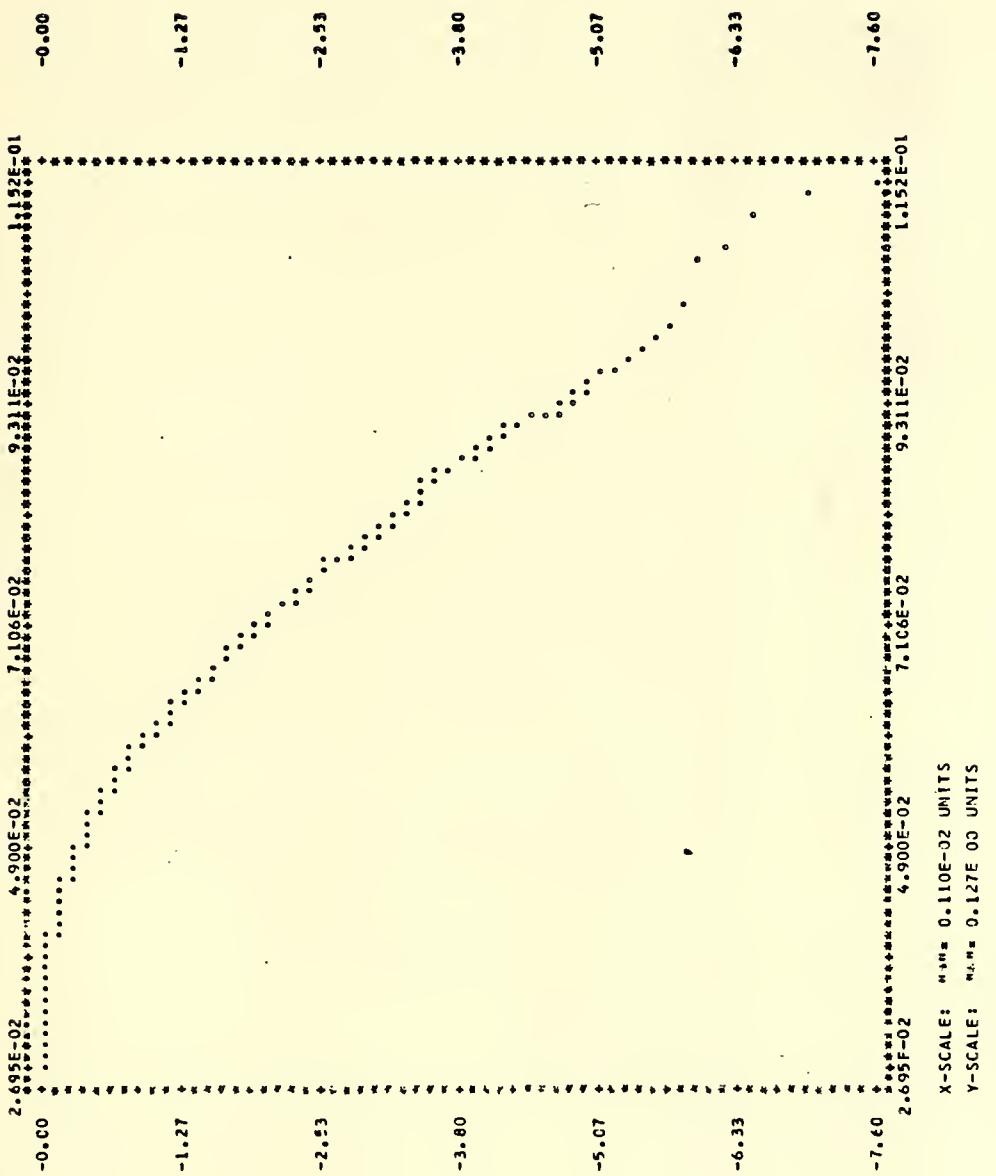
INTEGRAL SQUARE NORM SAMPLE SIZE M = 2000  
CAUCHY RANDOM VARIABLE SAMPLE SIZE, N = 500  
TRIANGULAR WINDOW. BANDWIDTH = 3/SQRT(M)



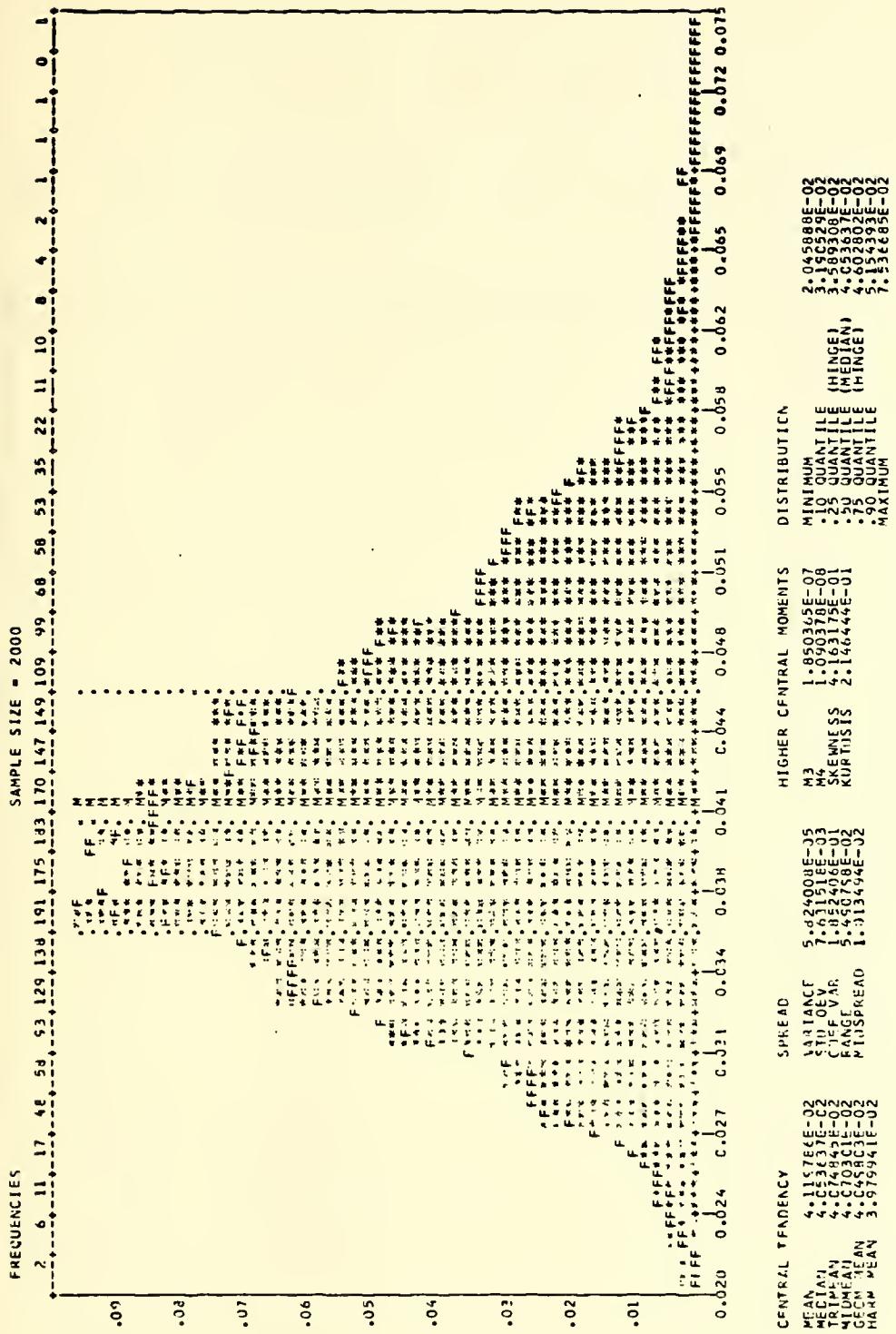


ESTIMATED CFNSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED OENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS









CENTRAL FREQUENCY

MEAN	0.11576E-02
MEDIAN	4.15363E-02
TRIMMED MEAN	4.05444E-02
TRILOGM. MEAN	4.05034E-02
GEOM. MEAN	4.05584E-02
HARM. MEAN	3.97594E-02

SPREAD

VARIANCE	5.02900E-35
STDEV	7.05110E-35
RANGE	1.02266E-01
KURTOSIS	5.446158E-01
SKURTOSIS	1.013444E-02
MJSPREAD	0.027

HIGHER CENTRAL MOMENTS

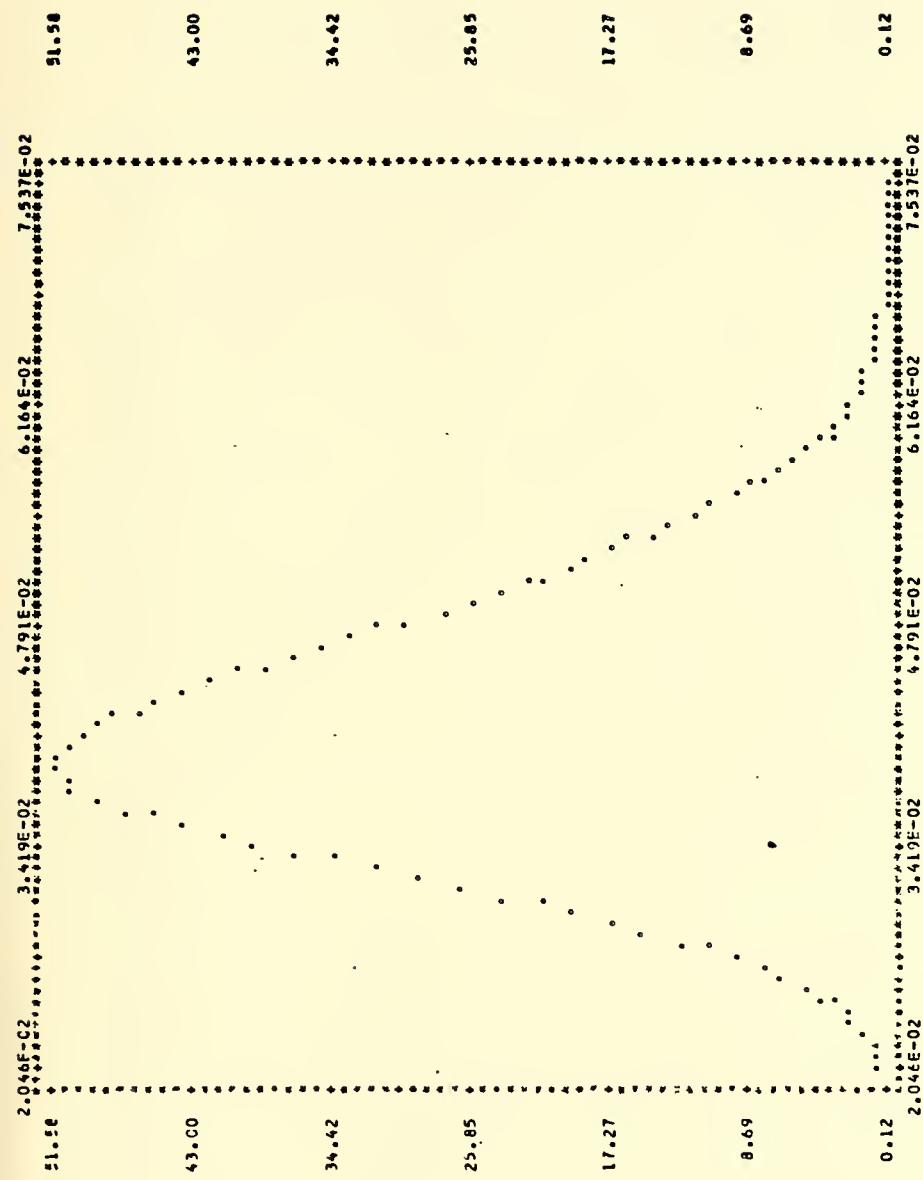
M3	1.050345E-07
M4	1.050371E-08
SKEWNESS	4.62317E-01
KURTOSIS	2.146444E-01

DISTRIBUTION

MINIMUM	1.050345E-07
1.050371E-08	1.050371E-08
2.050345E-02	2.050345E-02
3.050345E-02	3.050345E-02
4.050345E-02	4.050345E-02
5.050345E-02	5.050345E-02
6.050345E-02	6.050345E-02
7.050345E-02	7.050345E-02

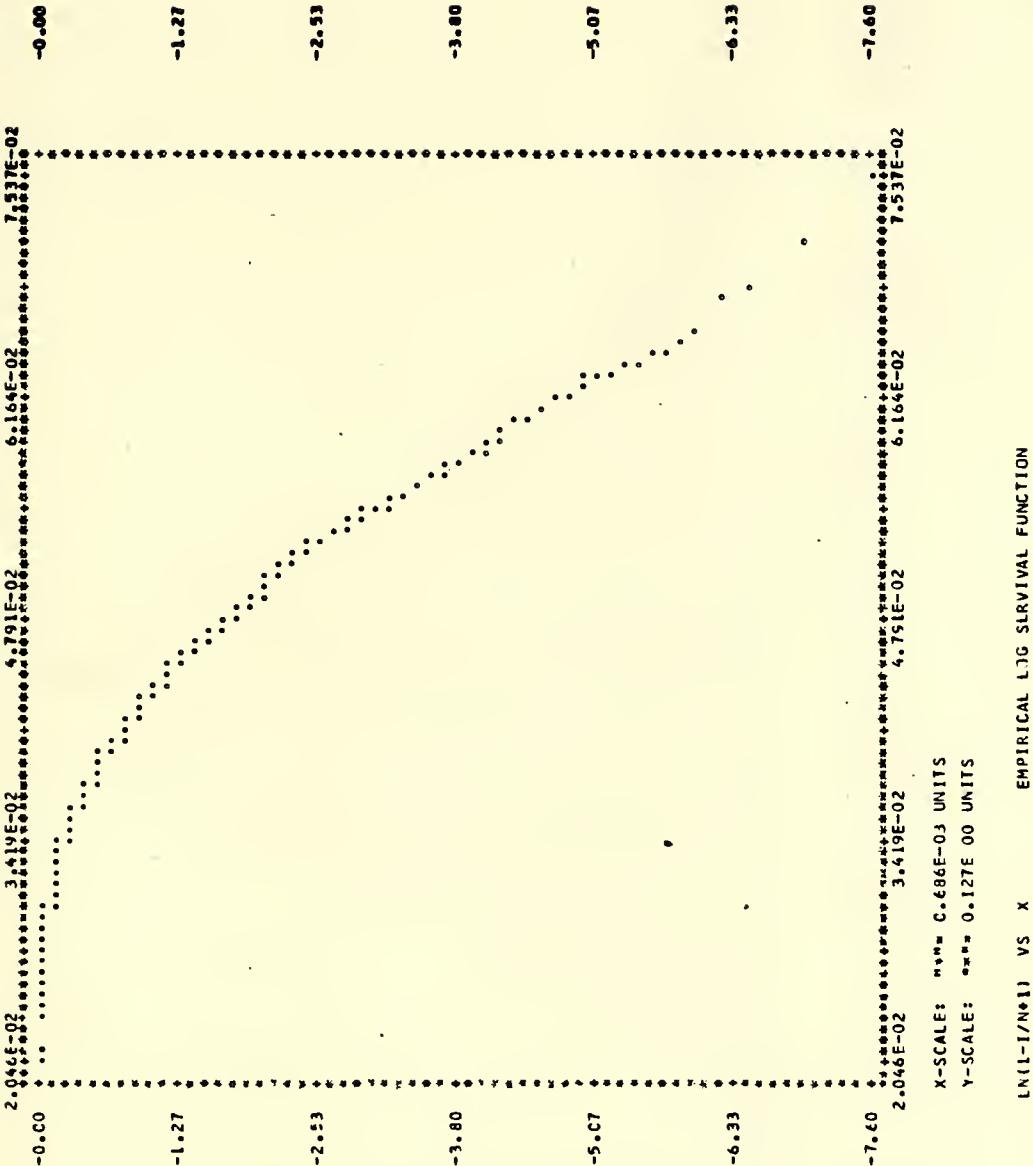
INTEGRAL SQUARE NORM SAMPLE SIZE  $n = 2000$   
 CAUCHY RANGE VARIABLE SAMPLE SIZE  $n = 1000$   
 TRIANGULAR WINDOW DENSITY = 3/SURTRN



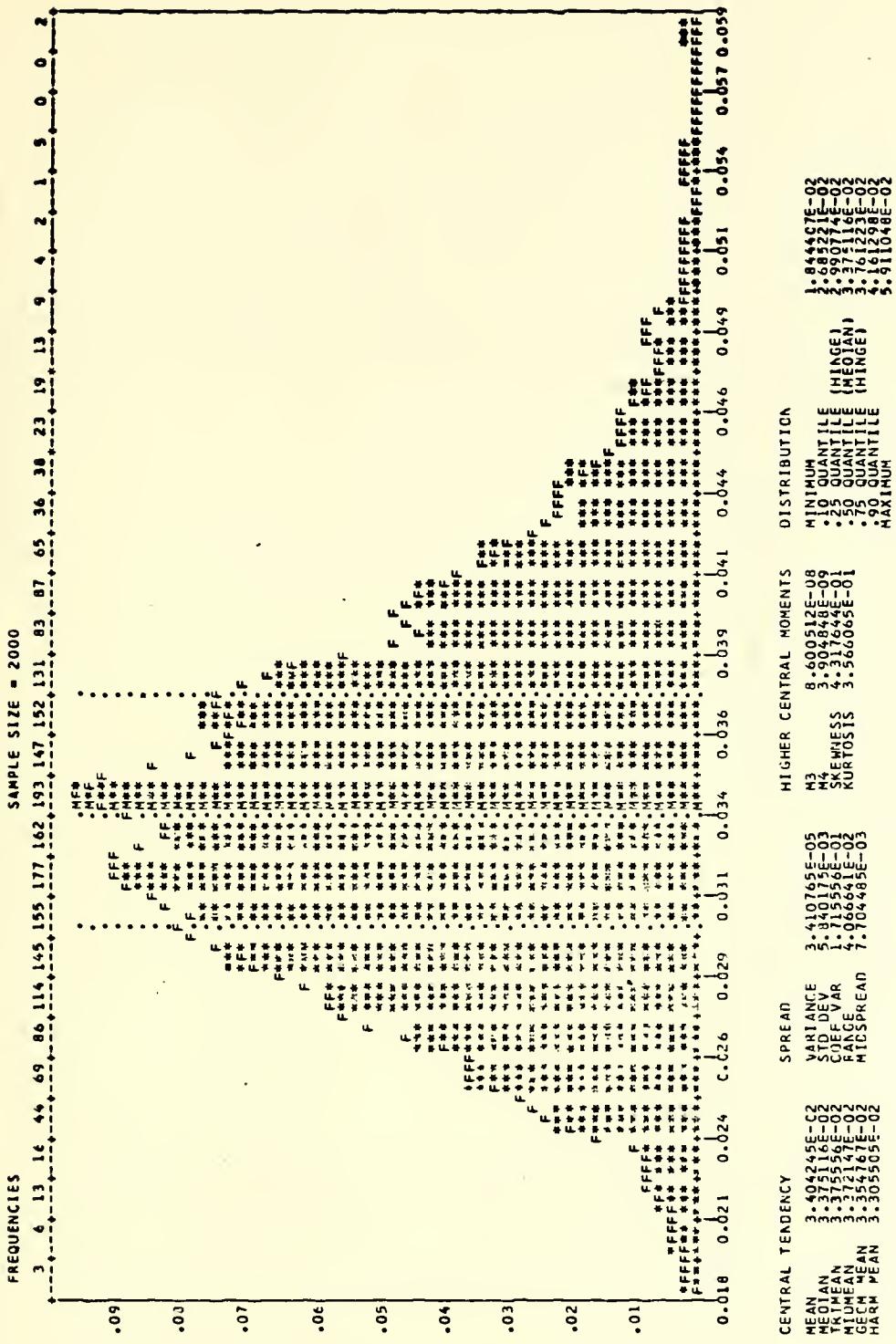


ESTIMATED CENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2200 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED CENSITY FUNCTION IS EVALUATE AT 100 EQUALLY SPACED POINTS



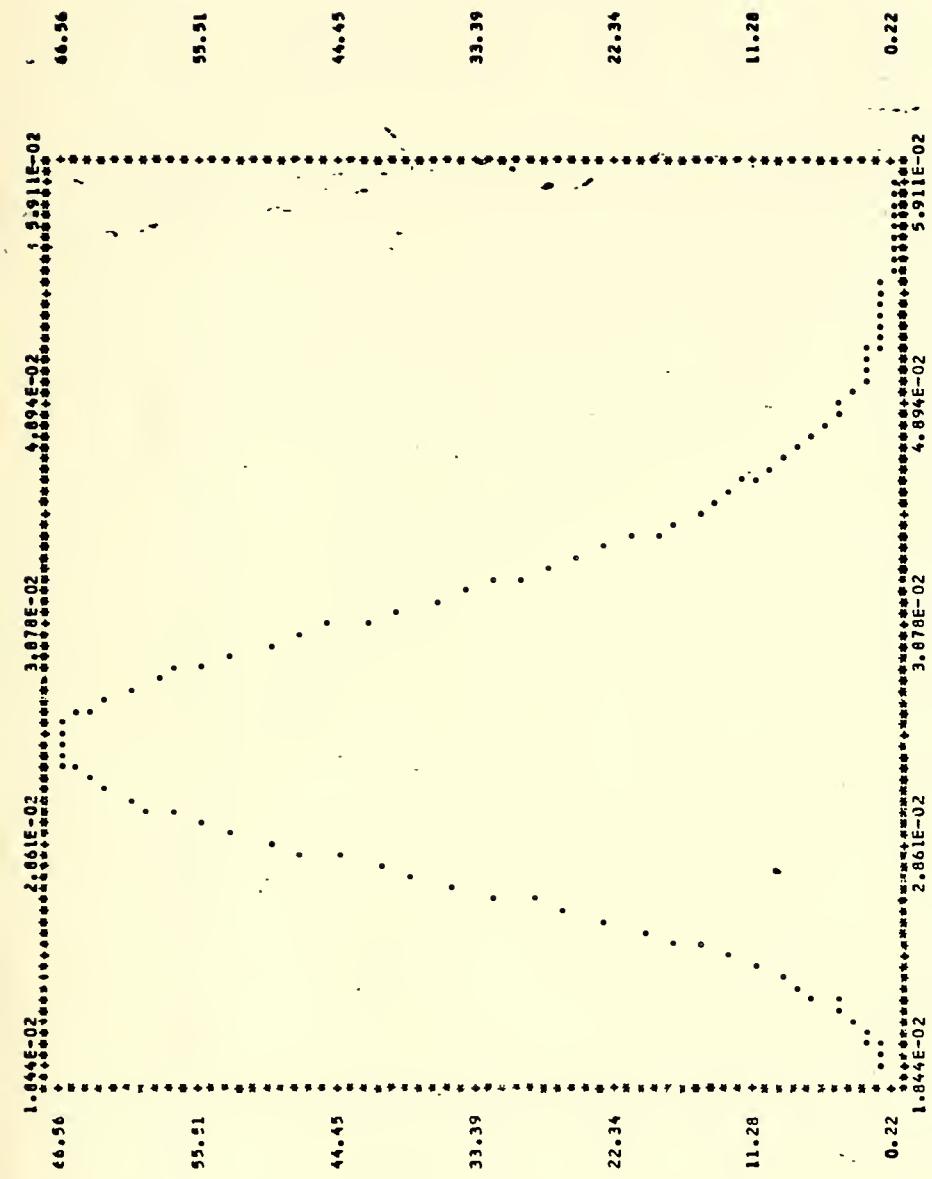






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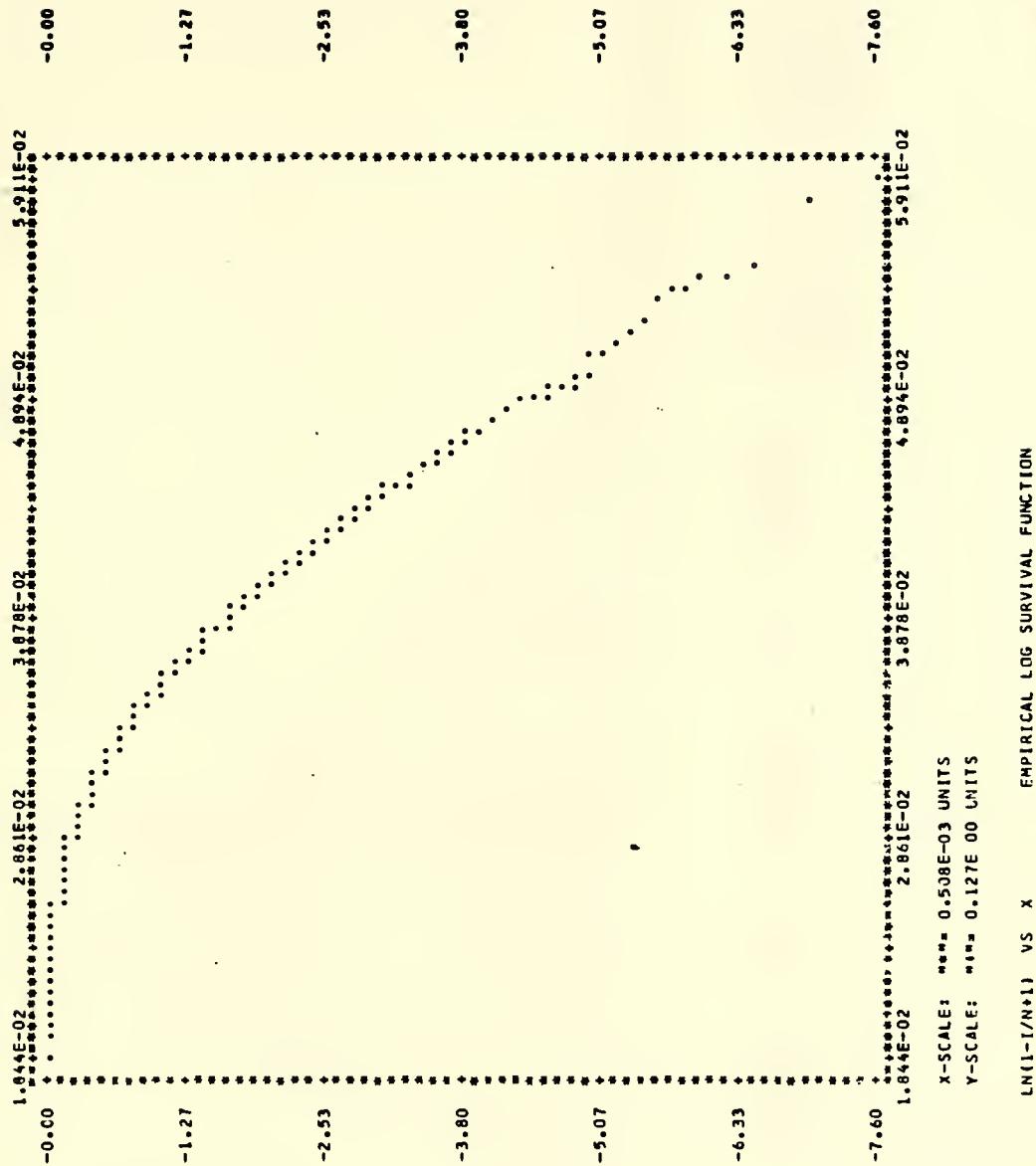




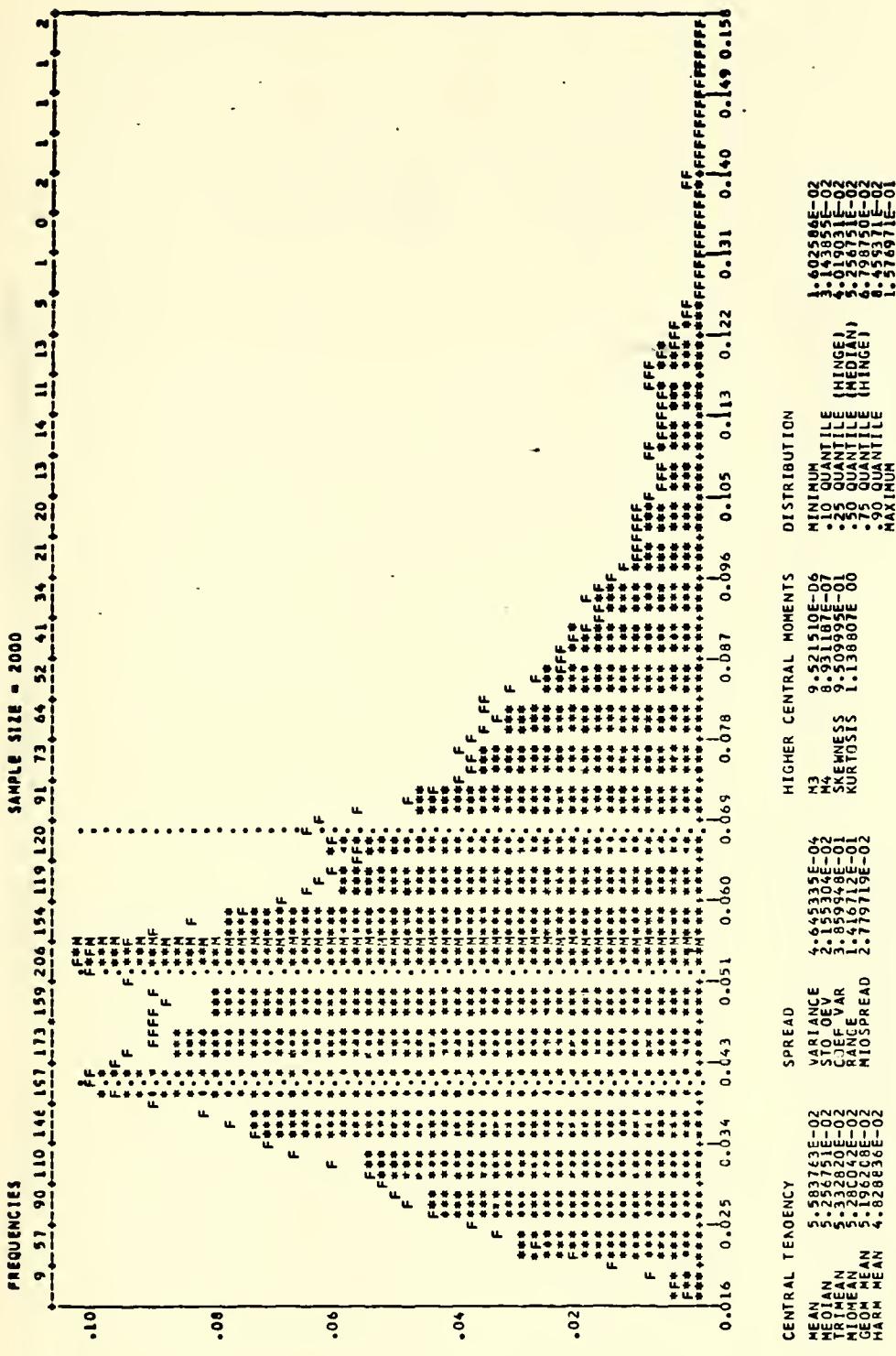
X-SCALE: \*\*\* = 0.5CBE-03 UNITS  
 Y-SCALE: \*\*\* = 0.111E 01 UNITS

ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NCRM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED CENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS



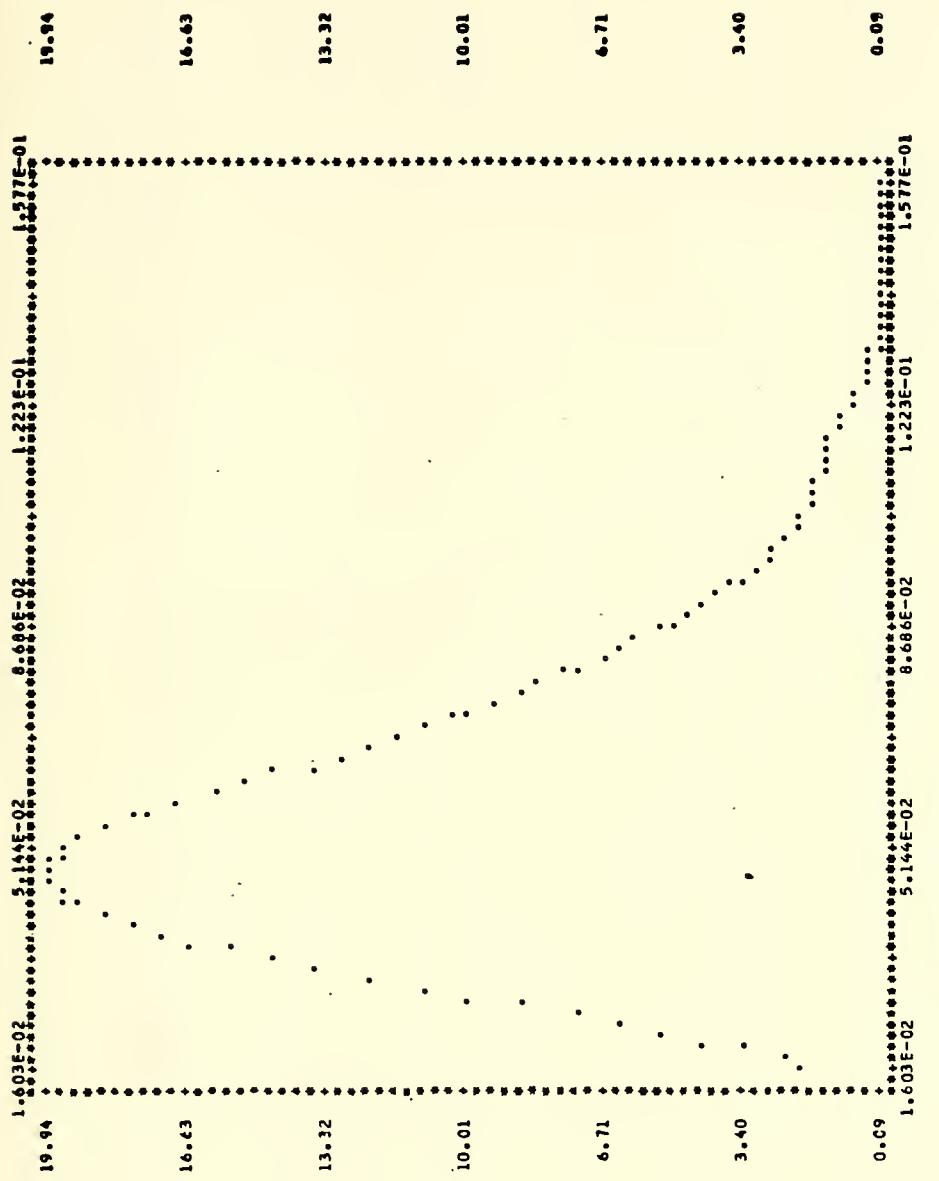






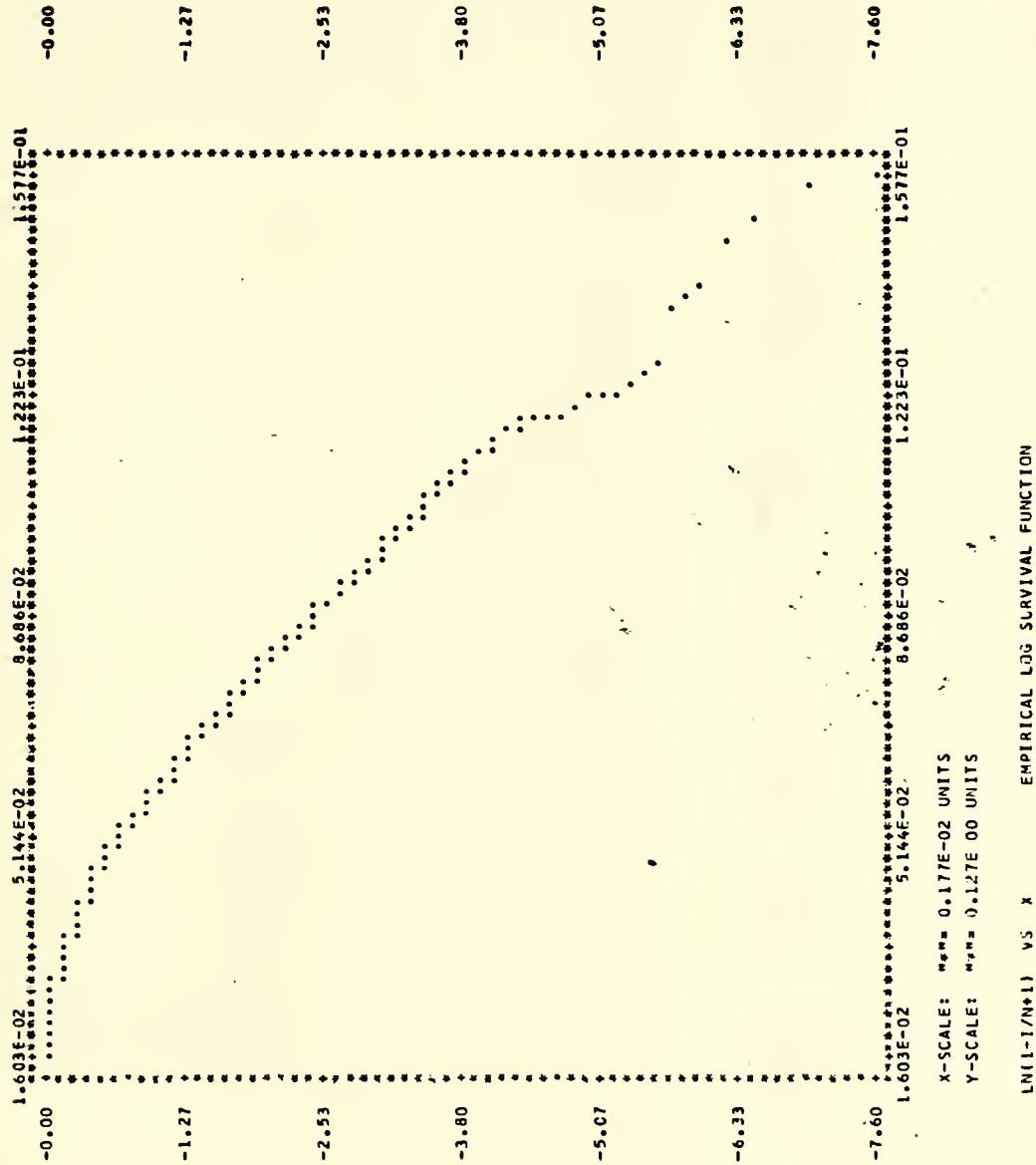
INTEGRAL SQUARE NORM SAMPLE SIZE N = 2000  
CAUCHY RANDOM VARIABLE SAMPLE SIZE N = 100  
TRIANGULAR MINONW. BANDWIDTH = 20/SQRT(N)



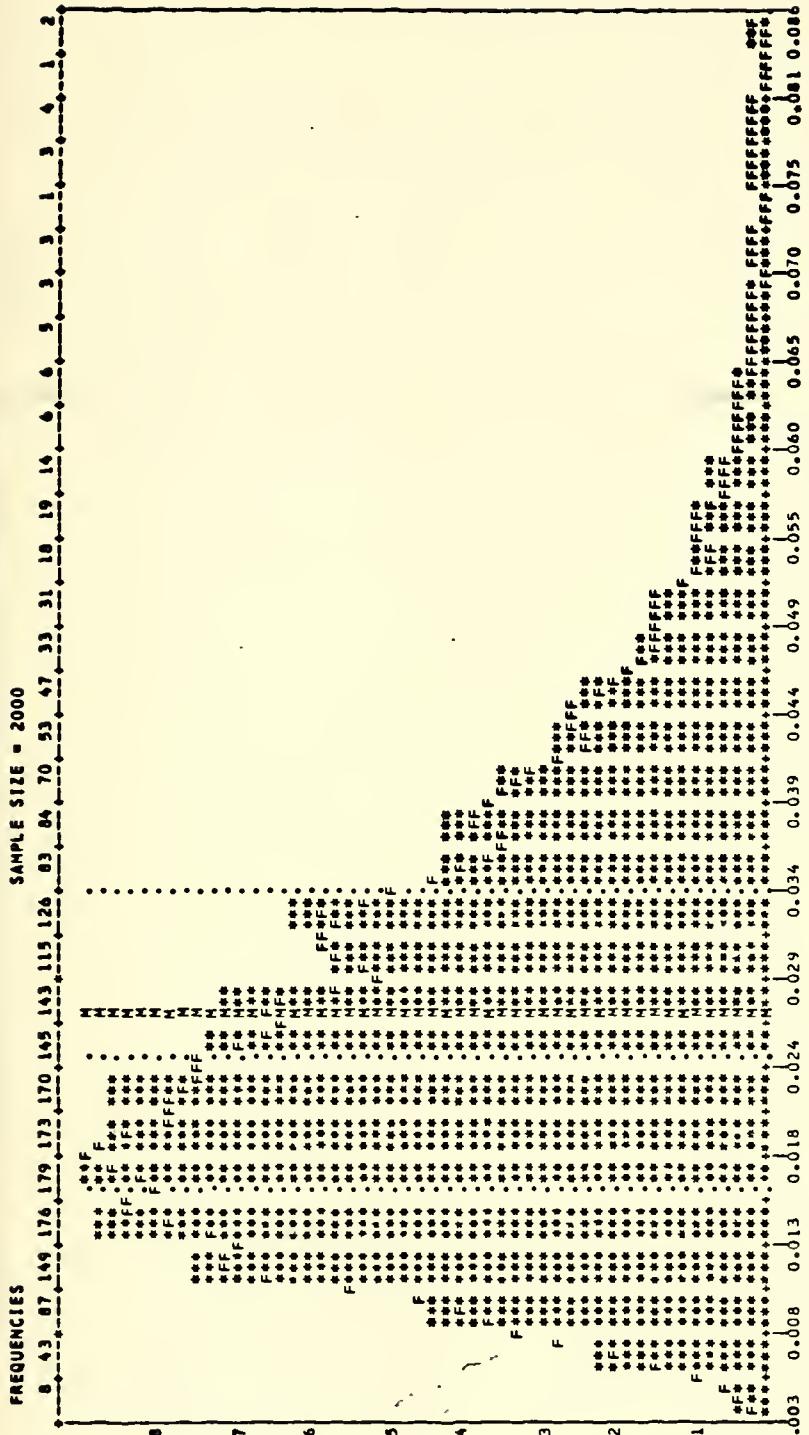


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000    TRIANGULAR WINDOW    BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS





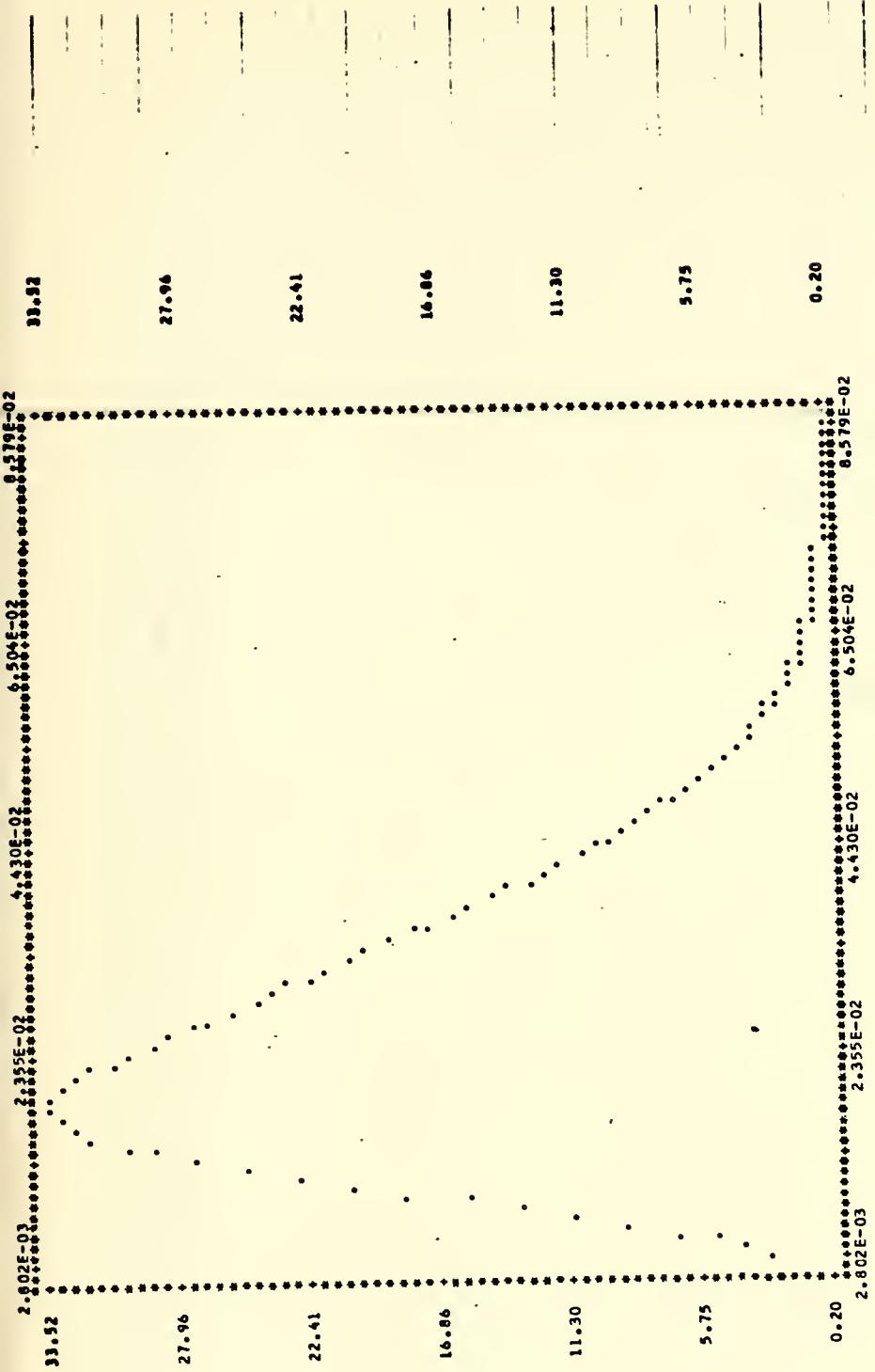




CENTRAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS			DISTRIBUTION
		M3	M4	MINIMUM	
MEAN	2.627237E-02	1.08773E-04	2.220775E-06	2.802986E-03	
MEOTIAN	2.378632E-02	1.107205E-02	1.225009E-07	1.166754E-02	
TRIMEAN	2.436662E-02	4.98319E-01	9.94206E-01	25 QUANTILE (HNGE)	
MOOMEAN	2.421955E-02	8.58866E-02	1.195372E-00	50 QUANTILE (MEOTIAN)	
GEOM MEAN	2.313033E-02	1.736961E-02		75 QUANTILE (HNGE)	
HARM MEAN	2.013422E-02			90 QUANTILE (HNGE)	
				MAXIMUM	8.57911E-02

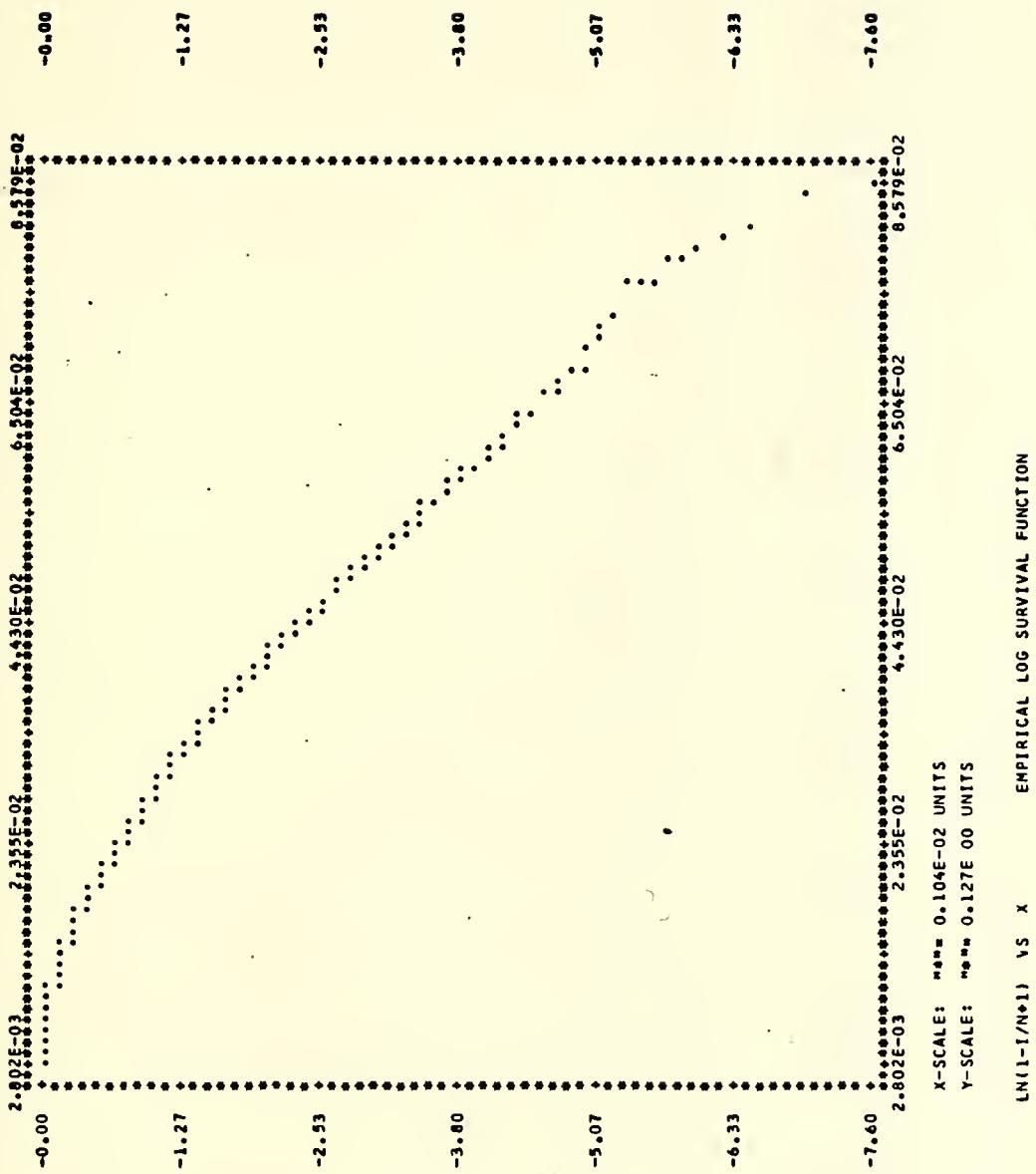
INTEGRAL SQUARE NORM SAMPLE SIZE M = 2000  
 CAUCHY RANDOM VARIABLE SAMPLE SIZE N = 200  
 TRIANGULAR WINDOW. BANDWIDTH = 20/SQR(TINI)



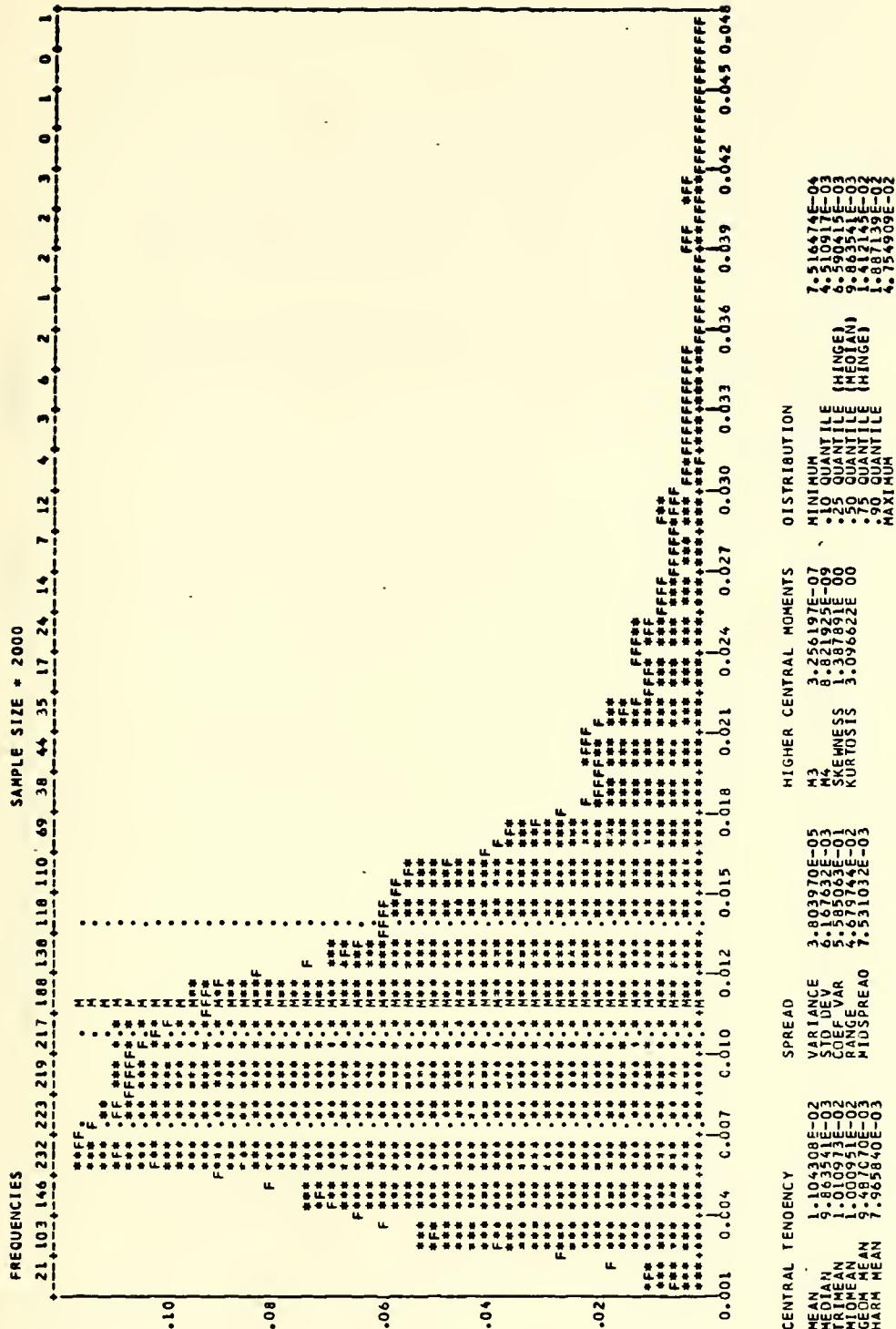


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 10  
 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS



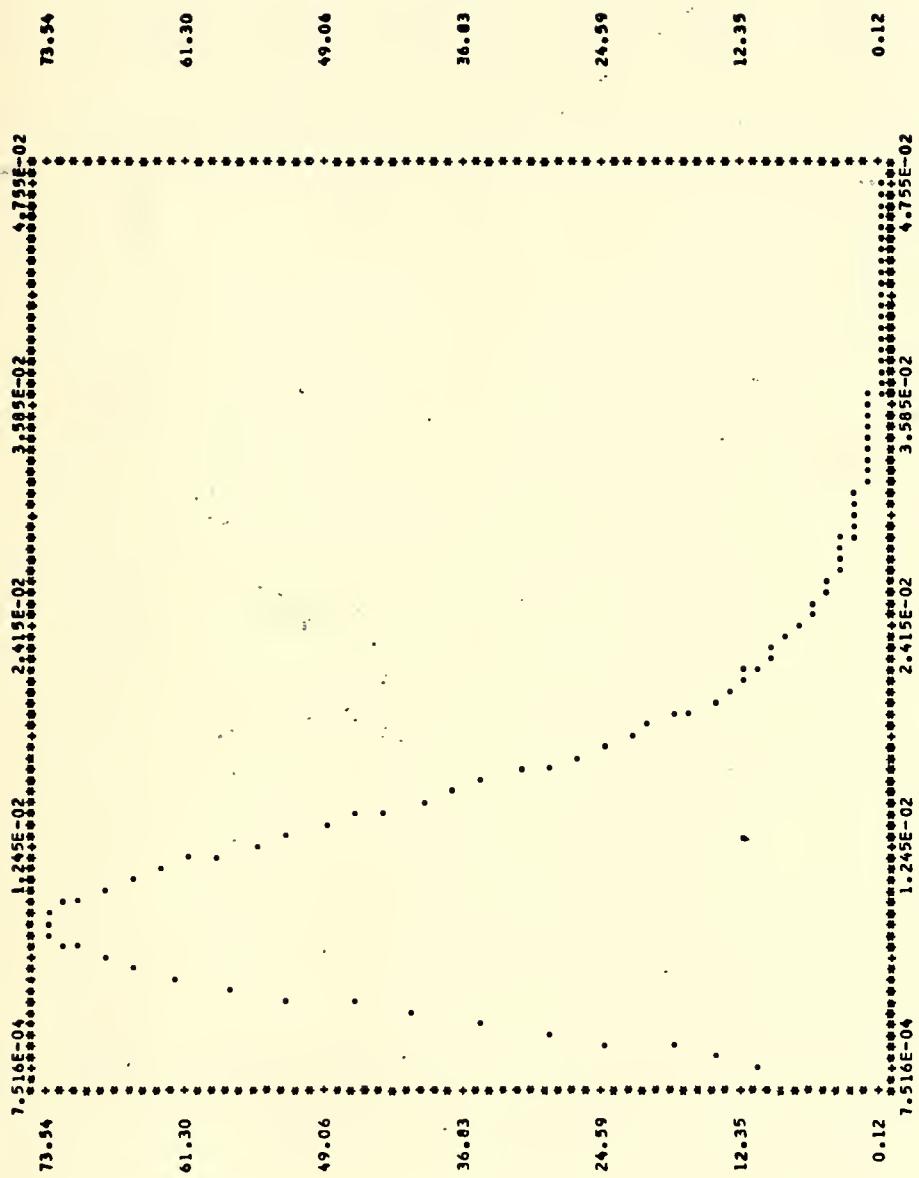






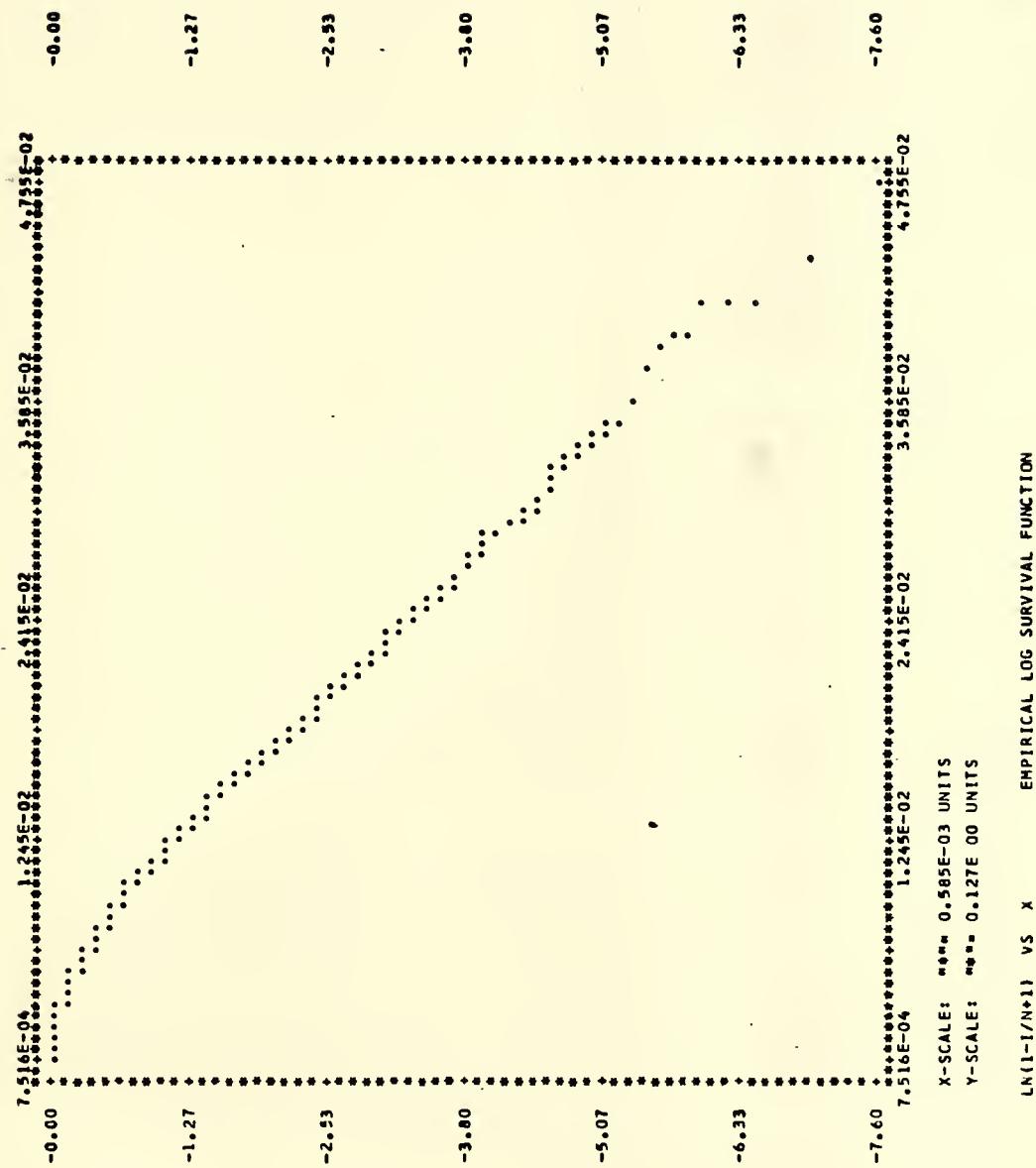
INTEGRAL SQUARE NORM SAMPLE SIZE M = 2000  
 CAUCHY RANDOM VARIABLE SAMPLE SIZE N = 500  
 TRIANGULAR WINDOW. BAOWIOTH = 20/SQRT(N)



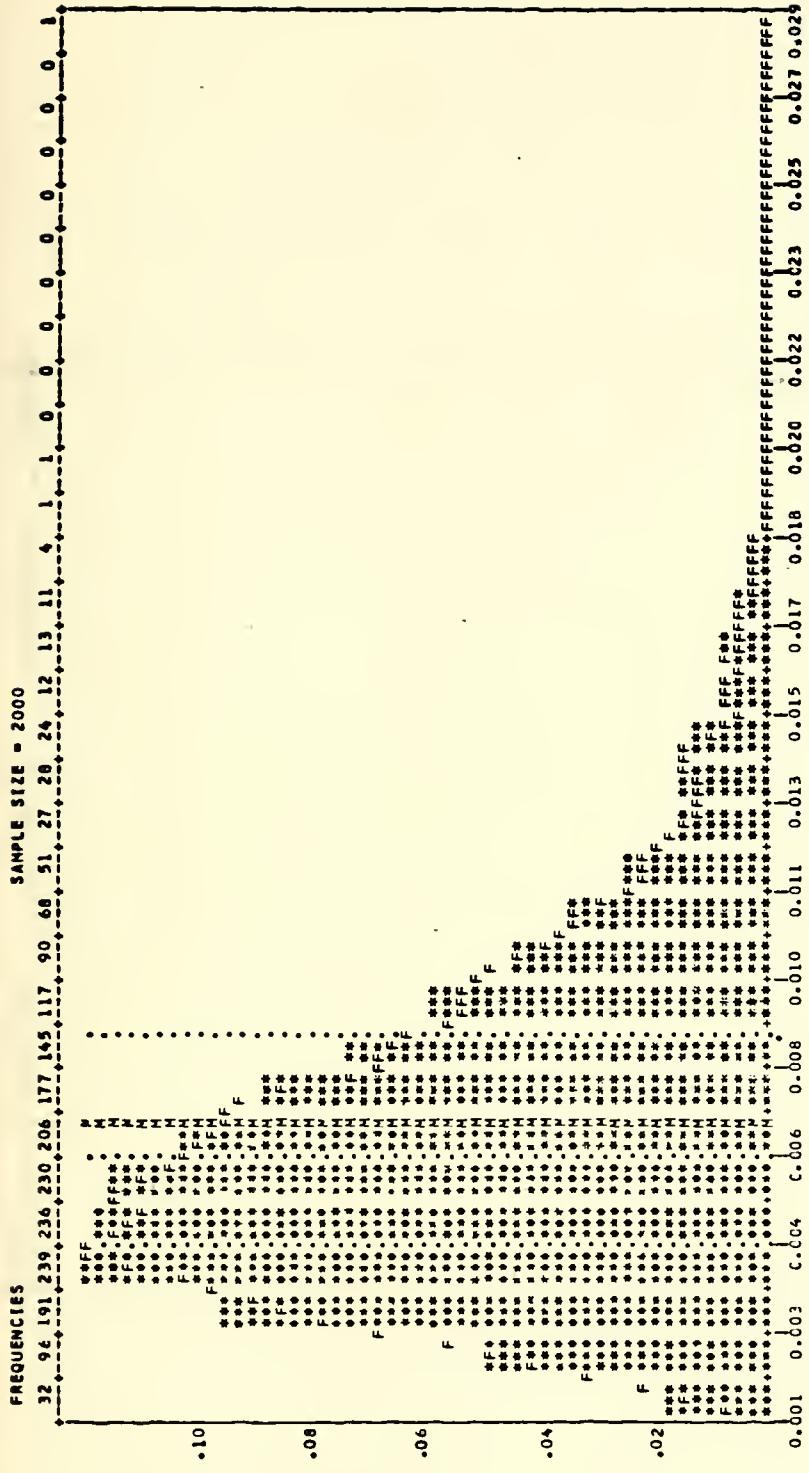


ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
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 ESTIMATED DENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS





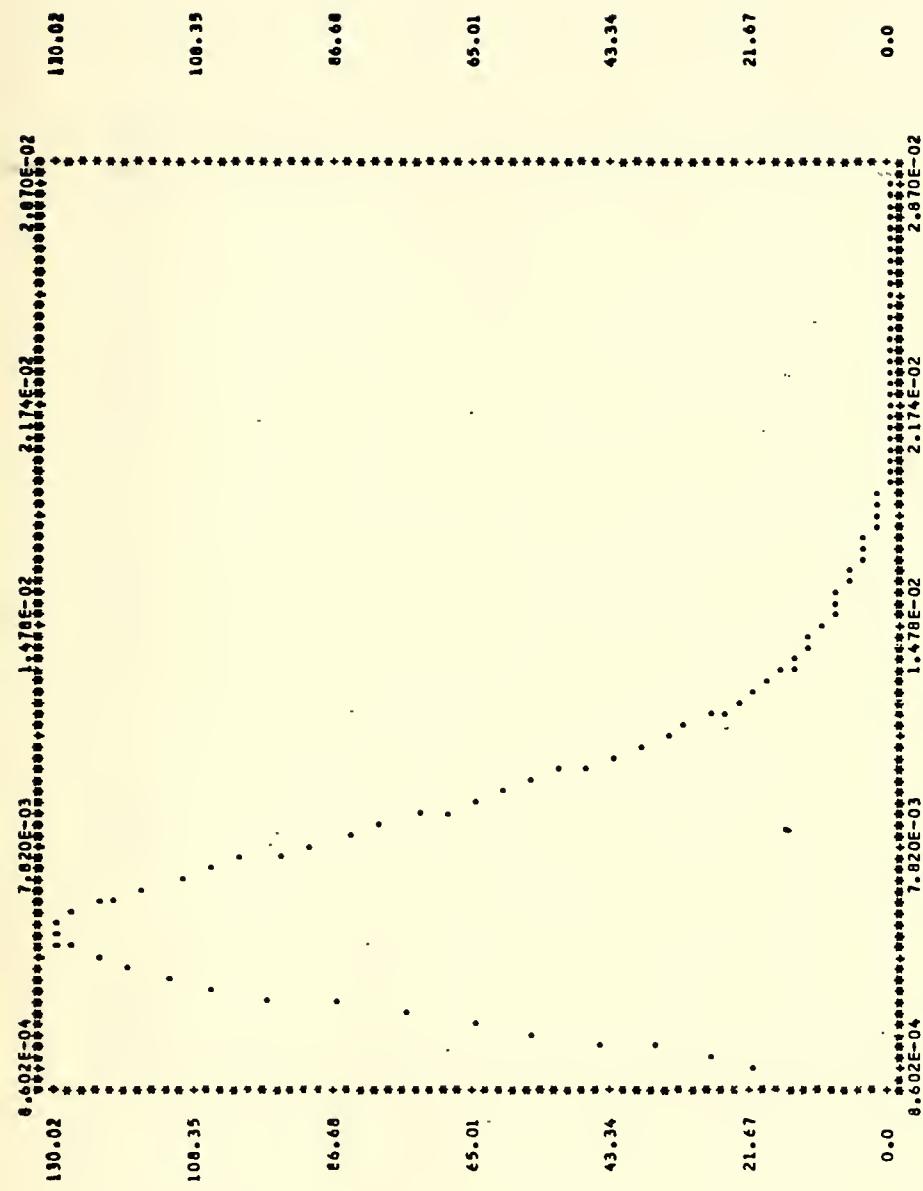




CENTRAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS				DISTRIBUTION
		VARIANCE	STD DEV	M <sup>3</sup>	M <sup>4</sup>	
MEAN	6.554537E-03	1.049055E-05	3.50936E-08	MINIMUM	3.50936E-08	MINIMUM
MEDIAN	6.584481E-03	1.049055E-05	5.101948E-10	0 QUANTILE	5.101948E-10	0 QUANTILE
MEAN	6.115555E-03	1.049055E-05	4.941162E-01	25 QUANTILE	4.941162E-01	25 QUANTILE
TRIMMED MEAN	6.115555E-03	1.049055E-05	2.784604E-02	50 QUANTILE	2.784604E-02	50 QUANTILE
MINIMUM	6.052288E-03	1.049055E-05	1.635954E-01	75 QUANTILE	1.635954E-01	75 QUANTILE
MAXIMUM	6.794682E-03	1.049055E-05	4.170049E-03	90 QUANTILE	4.170049E-03	90 QUANTILE
GEOMEAN	5.794682E-03	1.049055E-05	1.635954E-01	MAXIMUM	1.635954E-01	MAXIMUM
HARM. MEAN	5.046826E-03	1.049055E-05	4.170049E-03	MINIMUM	4.170049E-03	MINIMUM

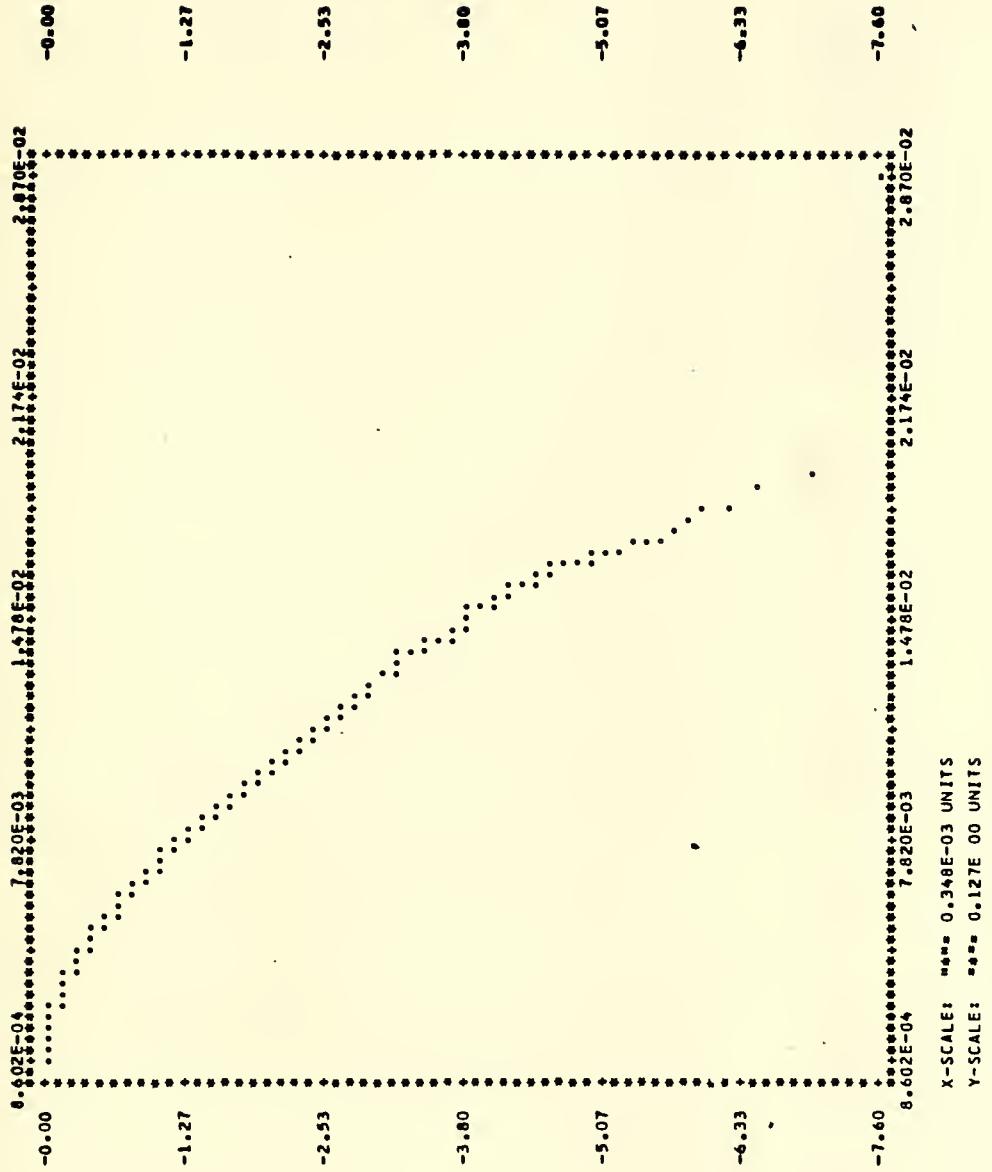
INTEGRAL SQUARE NORM SAMPLE SIZE  $n = 2000$   
 CAUCHY RANDOM VARIABLE SAMPLE SIZE  $n = 1000$   
 $\text{BANWOTH} = 20/\text{SQR}(n)$   
 TRIANGULAR WINDOW.





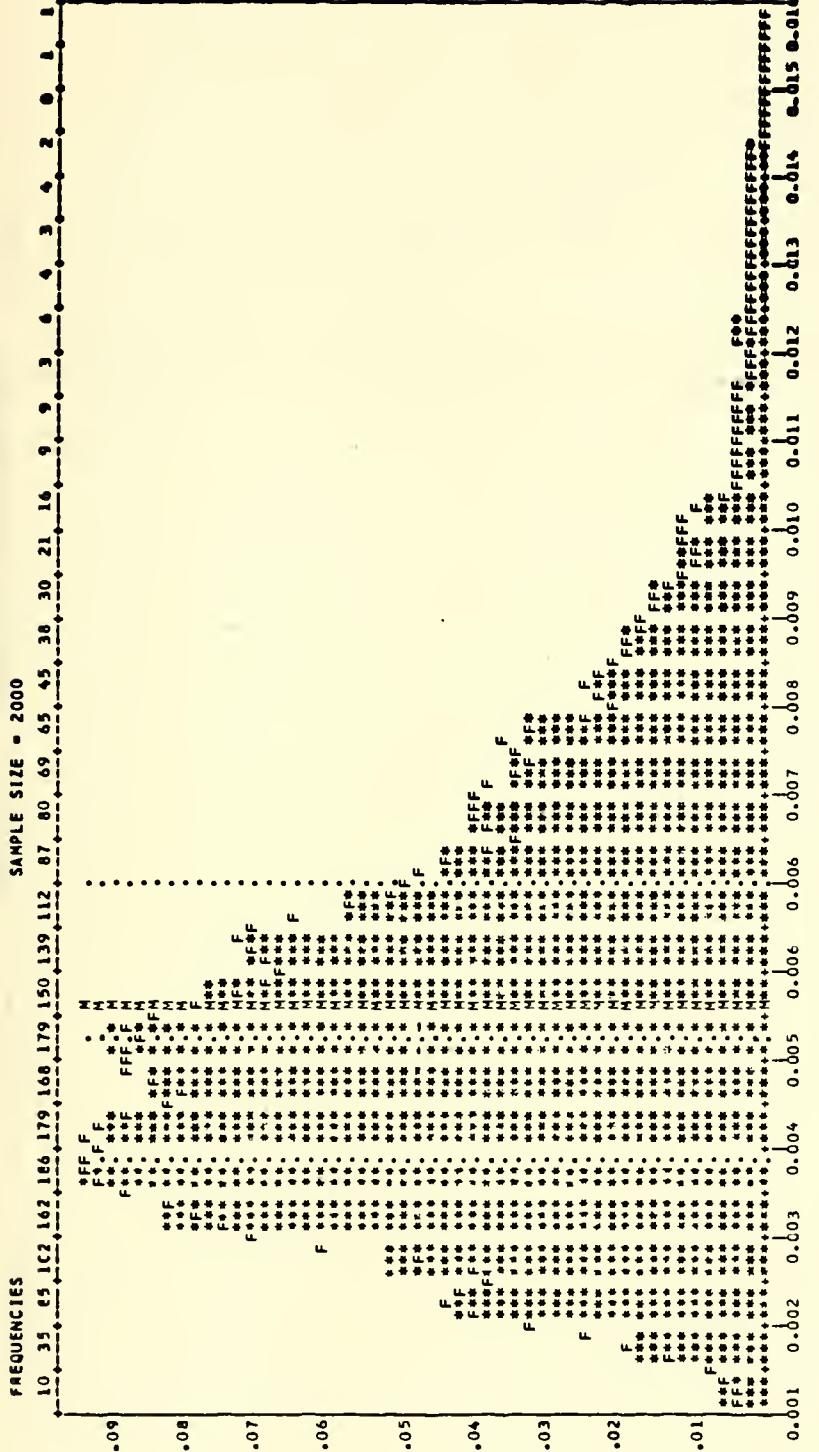
ESTIMATED CENSITY FUNCTION VS INTEGRAL SQUARE NORM RANDOM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000      TRIANGULAR WINDOW      BANOWIOTH = 1.0  
 ESTIMATED CENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS





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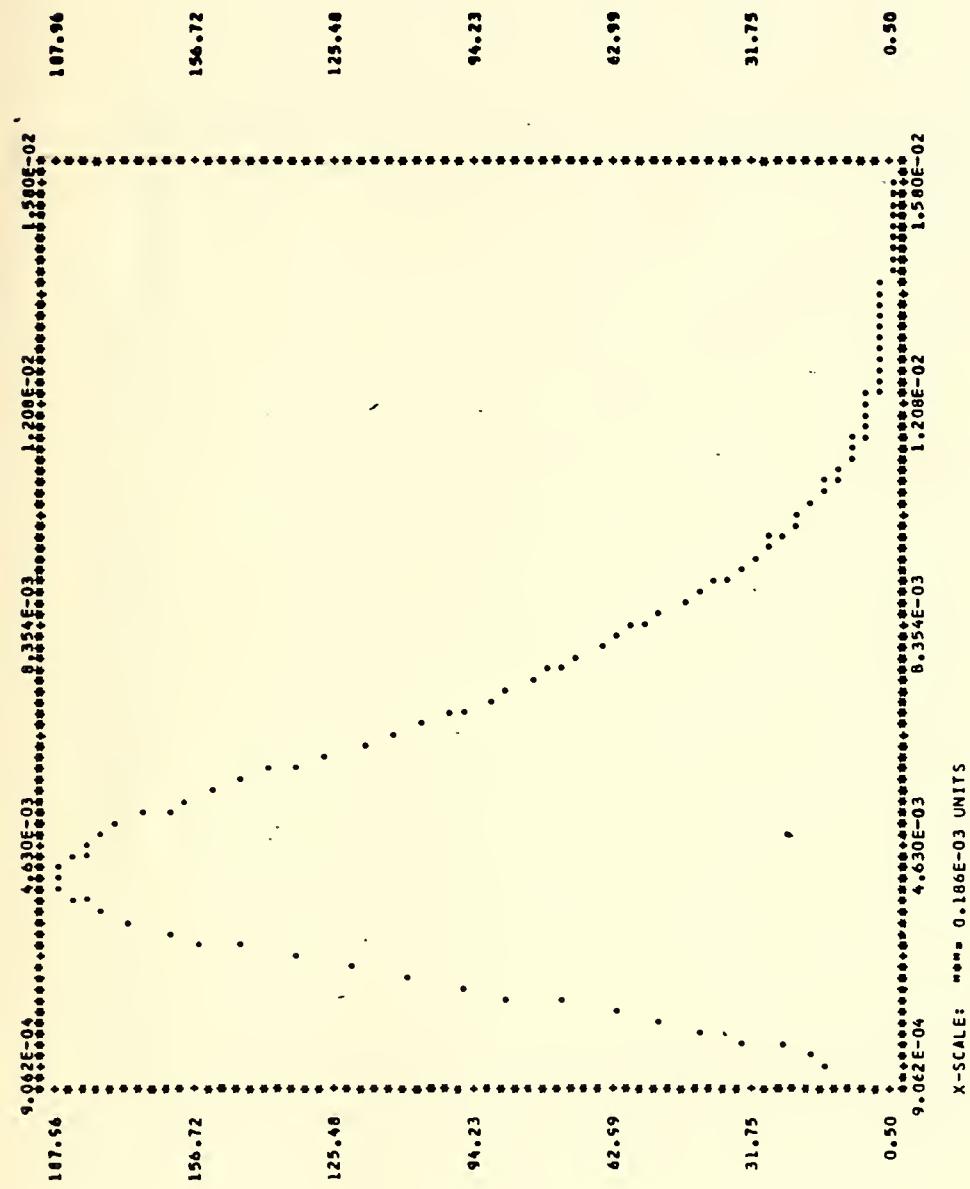




CENTRAL TENDENCY	SPREAD	HIGHER CENTRAL MOMENTS			DISTRIBUTION
		M3	M4	MINIMUM	
MEAN	5.1866167E-03	5.1336751E-03	1.0650100E-08	1.060388E-10	9.061934E-04
MEDIAN	4.8022433E-03	2.266444E-03	1.060388E-10	9.15226E-01	2.66321E-03
TRIMEAN	4.900172E-03	4.369577E-03	1.0185222E-00	1.0185222E-00	3.502941E-03
MODEAN	4.861325E-03	1.89953E-02	KURTOSIS	50 QUANTILE (HINGE)	4.82233E-03
GEOM MEAN	4.71068E-03	2.949947E-03		75 QUANTILE (HINGE)	6.452088E-03
HARM MEAN	4.24958E-C3			90 QUANTILE (HINGE)	8.20424E-03
				MAXIMUM	1.580156E-02

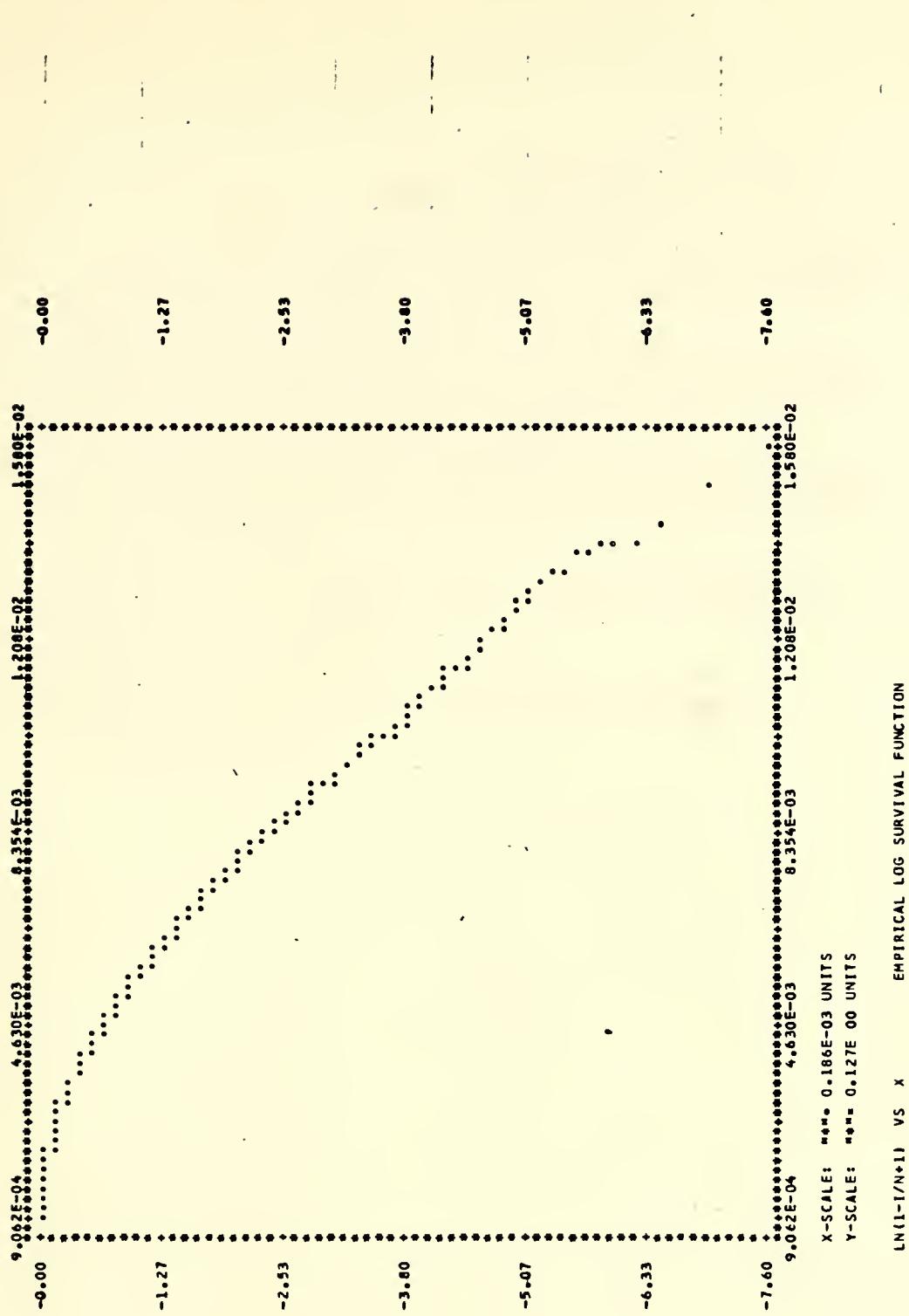
INTEGRAL SQUARE NORM SAMPLE SIZE M = 2000  
 CAUCHY RANDOM VARIABLE SAMPLE SIZE N = 1500  
 TRIANGULAR WINDOW. RADIOMICHTH = 20/SQRT(N)





ESTIMATED DENSITY FUNCTION VS INTEGRAL SQUARE NORM VARIABLE  
 INTEGRAL SQUARE NORM SAMPLE SIZE = 2000 TRIANGULAR WINDOW BANDWIDTH = RANGE / 1.0  
 ESTIMATED CENSITY FUNCTION IS EVALUATED AT 100 EQUALLY SPACED POINTS







#### LIST OF REFERENCES

1. Bartlett, M. S., (1963). Statistical estimation of density functions. *Sankhyā, Ser. A*25, 245-254.
2. Bickel, P. J. and Rosenblatt, M., (1973). On some global measures of the deviations of density function estimates. *The Annals of Mathematical Statistics*, I, 1071-1095.
3. Rosenblatt, M., (1956). Remarks on some non-parametric estimates of a density function. *The Annals of Mathematical Statistics*, 27, 3.
4. Rosenblatt, M., (1971). Curve estimates. *The Annals of Mathematical Statistics*, 42, No. 6.
5. Wegman, E. J., (1972). Non-parametric probability density estimation: I. A summary of available methods. *Technometrics*, 14, No. 3.



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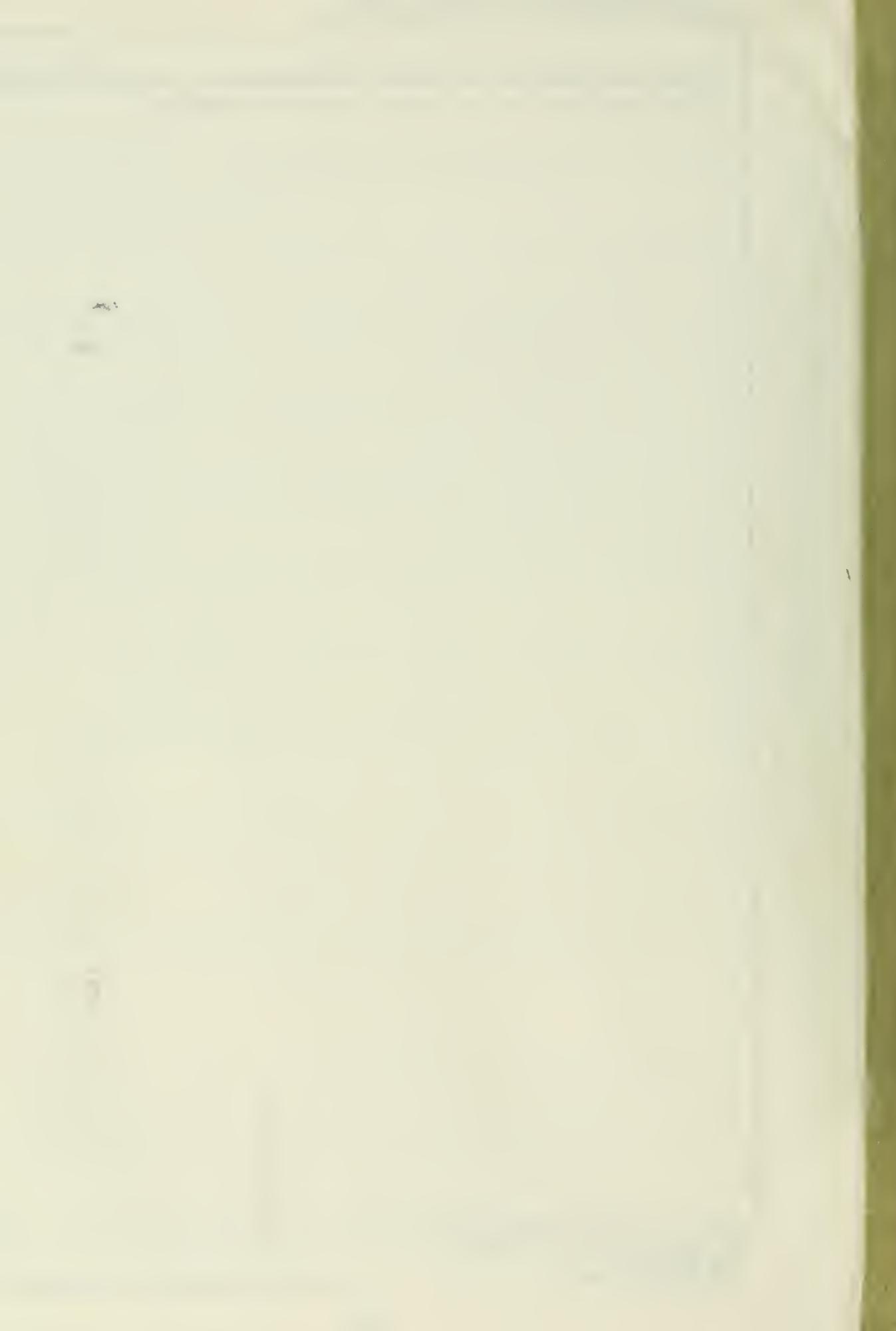


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has not been possible to relate the parameters to the population distribution characteristics or to the window shape and bandwidth.

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