

Modified Box-Cox Transformation and Manly Transformation with Failure Time Data

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Backgrounds

- ▶ Many procedures require data to be approximately normal.
- ▶ A transformation that transforms the data set to achieve normality is used.



Data Transformation

- ▶ Based on the relationship between the standard deviation and the mean

Relationship between σ and μ	Transformation
$\sigma \propto \mu^{1/2}$	Square root
$\sigma \propto \mu$	Logarithmic
$\sigma \propto \mu^{3/2}$	Reciprocal square root
$\sigma \propto \mu^2$	Reciprocal

Source: Montgomery, 2001: 84.



▶ Transformations for Specific Distributions

For Example, the square root transformation is used for Poisson data, the logarithmic transformation for lognormal data and the arcsine transformation for binomial data expressed as fractions.



▶ A family of transformations

Box and Cox (1964)

$$Y = \begin{cases} \frac{X^\lambda - 1}{\lambda}, & \lambda \neq 0 \\ \ln X & , \lambda = 0 \end{cases}$$



$$Y = \begin{cases} \frac{[X + c]^\lambda - 1}{\lambda} & , \lambda \neq 0 \\ \ln[X + c] & , \lambda = 0 \end{cases}$$

- ▶ where c is translation constant



Cautions for the Box-Cox Transformation

- ▶ John and Draper (1980) showed that the Box-Cox Transformation was not satisfactory even when the best value of transformation parameter had been chosen.
- ▶ Doksum and Wong (1983) indicated that the Box-Cox transformation should be used with caution in some cases such as failure time and survival data.



▶ **Schlesselman (1971)**

$$Y = \begin{cases} \frac{X^\lambda - c^\lambda}{\lambda}, & \lambda \neq 0 \\ \ln(X/c), & \lambda = 0 \end{cases}$$

- ▶ where c is an arbitrary positive constant in the measurement units of variable X .



► **Manly(1976)**

$$Y = \begin{cases} \frac{\exp(\lambda X) - 1}{\lambda}, & \lambda \neq 0 \\ X & , \lambda = 0 \end{cases}$$



The Modified Box and Cox transformation

► Yeo and Johnson (2000)

$$Y = \begin{cases} \frac{(X+1)^\lambda - 1}{\lambda} & , x \geq 0, \lambda \neq 0 \\ \ln(X+1) & , x \geq 0, \lambda = 0 \\ -\frac{(-X+1)^{2-\lambda} - 1}{2-\lambda} & , x < 0, \lambda \neq 0 \\ -\ln(-X+1) & , x < 0, \lambda = 0 \end{cases}$$



For Example

- ▶ Rahman M. and Pearson, L.M. (2007). A Note on the Modified Box-Cox Transformation. **Festschrift in honor of Distinguished Professor Mir Masoom Ali on the occasion of his retirement , May 18-19.** 106-115.
 - ▶ Abbasi, B., Niaki, S.T.A. and Seyedan, S.E. (2011). A Simple Transformation Method in Skewness Reduction. **IJE Transactions A:Basics.** 24(2): 169-175.
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Failure Time Data

- ▶ Gamma Distribution

$$f(x) = \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-\frac{x}{\beta}} & , x \geq 0, \alpha, \beta > 0 \\ 0 & , x < 0 \end{cases}$$



Failure Time Data

- ▶ Exponential distribution

$$f(x) = \begin{cases} \frac{1}{\beta} e^{-\left(\frac{x}{\beta}\right)} & , x \geq 0; \beta > 0 \\ 0 & , x < 0 \end{cases}$$



Comparisons of Several Population Means

- ▶ The probability density function of each transformed observation is in the form

$$f(y_{ij} | \mu_i, \sigma^2) = \frac{1}{(2\pi\sigma^2)^{1/2}} \exp\left\{-\frac{1}{2\sigma^2} (y_{ij} - \mu_i)^2\right\}$$



Estimation of Transformation Parameter for Modified Box and Cox Transformation

- ▶ The likelihood function in relation to the original observations is given by

$$L(\mu_i, \sigma^2, \lambda | x_{\bar{y}}) = \frac{1}{(2\pi\sigma^2)^{n/2}} \exp \left\{ -\frac{1}{2\sigma^2} \sum_{i=1}^k \sum_{j=1}^{n_i} \left[\frac{(x_{\bar{y}} + 1)^\lambda - 1}{\lambda} - \mu_i \right]^2 \right\} \cdot J(y^*, x) \quad \text{where}$$

$$J(y^*, x) = \prod_{i=1}^k \prod_{j=1}^{n_i} \left| \frac{\partial y_{\bar{y}}}{\partial x_{ij}} \right|.$$



Transformation Parameter in Modified Box -Cox transformation

- ▶ The maximum likelihood estimate of transformation parameter is obtained by solving the likelihood equation

$$\frac{d}{d\lambda} \ln L(\lambda) =$$

$$-n \left[\frac{\sum_{i=1}^k \sum_{j=1}^{n_i} (x_{ij} + 1)^{2\lambda} \ln(x_{ij} + 1) - \sum_{i=1}^k \frac{1}{n_i} \left(\sum_{j=1}^{n_i} (x_{ij} + 1)^\lambda \right) \left(\sum_{j=1}^{n_i} x_{ij}^\lambda \ln(x_{ij} + 1) \right)}{\sum_{i=1}^k \sum_{j=1}^{n_i} (x_{ij} + 1)^{2\lambda} - \sum_{i=1}^k \frac{1}{n_i} \left(\sum_{j=1}^{n_i} (x_{ij} + 1)^\lambda \right)^2} \right]$$

$$+ \frac{n}{\lambda} + \sum_{i=1}^k \sum_{j=1}^{n_i} \ln(x_{ij} + 1) = 0.$$



Estimation of Transformation Parameter for Manly Transformation

- ▶ The likelihood function in relation to the original observations is given by

$$L(\mu_i, \sigma^2, \lambda | x_{ij}) = \frac{1}{(2\pi\sigma^2)^{n/2}} \exp \left\{ -\frac{1}{2\sigma^2} \sum_{i=1}^k \sum_{j=1}^{n_i} \left[\frac{\exp(\lambda x_{ij}) - 1}{\lambda} - \mu_i \right]^2 \right\} \cdot J(y; x) \quad \text{where}$$

$$J(y; x) = \prod_{i=1}^k \prod_{j=1}^{n_i} \left| \frac{\partial y_{ij}}{\partial x_{ij}} \right|.$$



Transformation parameter in Manly Transformation

- ▶ The maximum likelihood estimate of transformation parameter is obtained by solving the likelihood equation

$$\frac{d}{d\lambda} \ln L(\lambda) =$$

$$-n \left[\sum_{i=1}^k \sum_{j=1}^{n_i} e^{2\lambda x_{ij}} x_{ij} - \sum_{i=1}^k \frac{1}{n_i} \left(\sum_{j=1}^{n_i} e^{\lambda x_{ij}} \right) \left(\sum_{j=1}^{n_i} e^{\lambda x_{ij}} x_{ij} \right) \right]$$

$$\sum_{i=1}^k \sum_{j=1}^{n_i} e^{2\lambda x_{ij}} - \sum_{i=1}^k \frac{1}{n_i} \left(\sum_{j=1}^{n_i} e^{\lambda x_{ij}} \right)^2$$

$$+ \frac{n}{\lambda} + \sum_{i=1}^k \sum_{j=1}^{n_i} x_{ij} = 0.$$

Check Validity of Assumption

- ▶ The Kolmogorov-Smirnov Test
- ▶ The Levene Test



SIMULATION STUDY FOR THE GAMMA DATA

▶ The possible value for study is set as follows:

k = number of the Gamma populations = 3,

n_i = sample size from the i th Gamma population is between 5 and 90,

β_i = scale parameter of the i th Gamma population is between 1 and 3 ,

α_i = shape parameter of the i th Gamma population is between 1 and 5

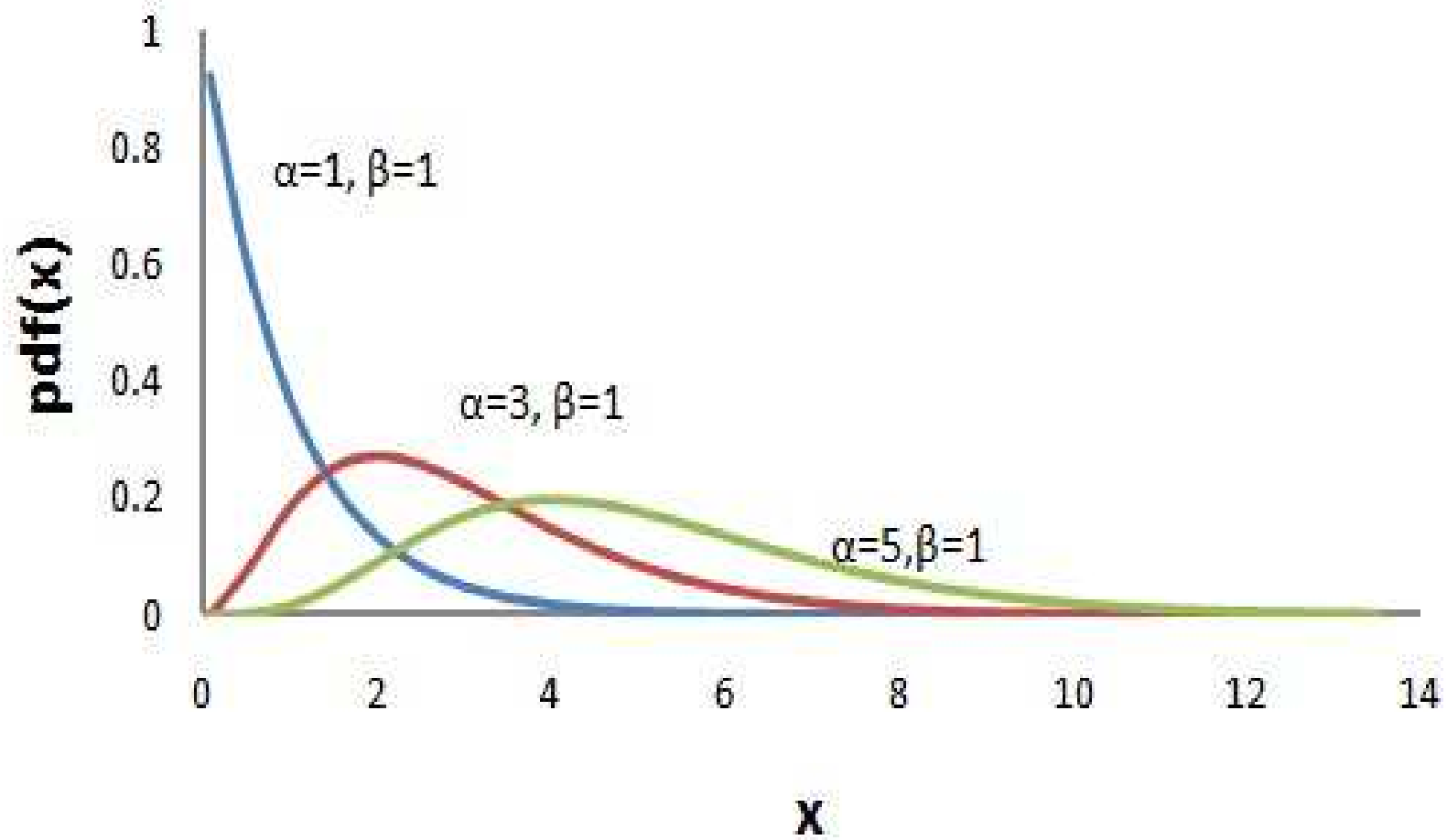


The Results

- ▶ The results of the goodness- of-fit tests and the tests of homogeneity of variances with 1,000 replicated samples of various sizes are as follows



Graph of Gamma Distribution



Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=10$	Manly	0.7861	0.7858	0.7933
	Modified	0.7866	0.7883	0.7910
$n_i=30$	Manly	0.6245	0.6563	0.6958
	Modified	0.6262	0.6640	0.6930



Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=50$	Manly	0.5045	0.5427	0.5975
	Modified	0.5077	0.5558	0.5930
$n_i=80$	Manly	0.3625	0.4124	0.4799
	Modified	0.3656	0.4294	0.4904



Averages of the p-Values for K-S Test

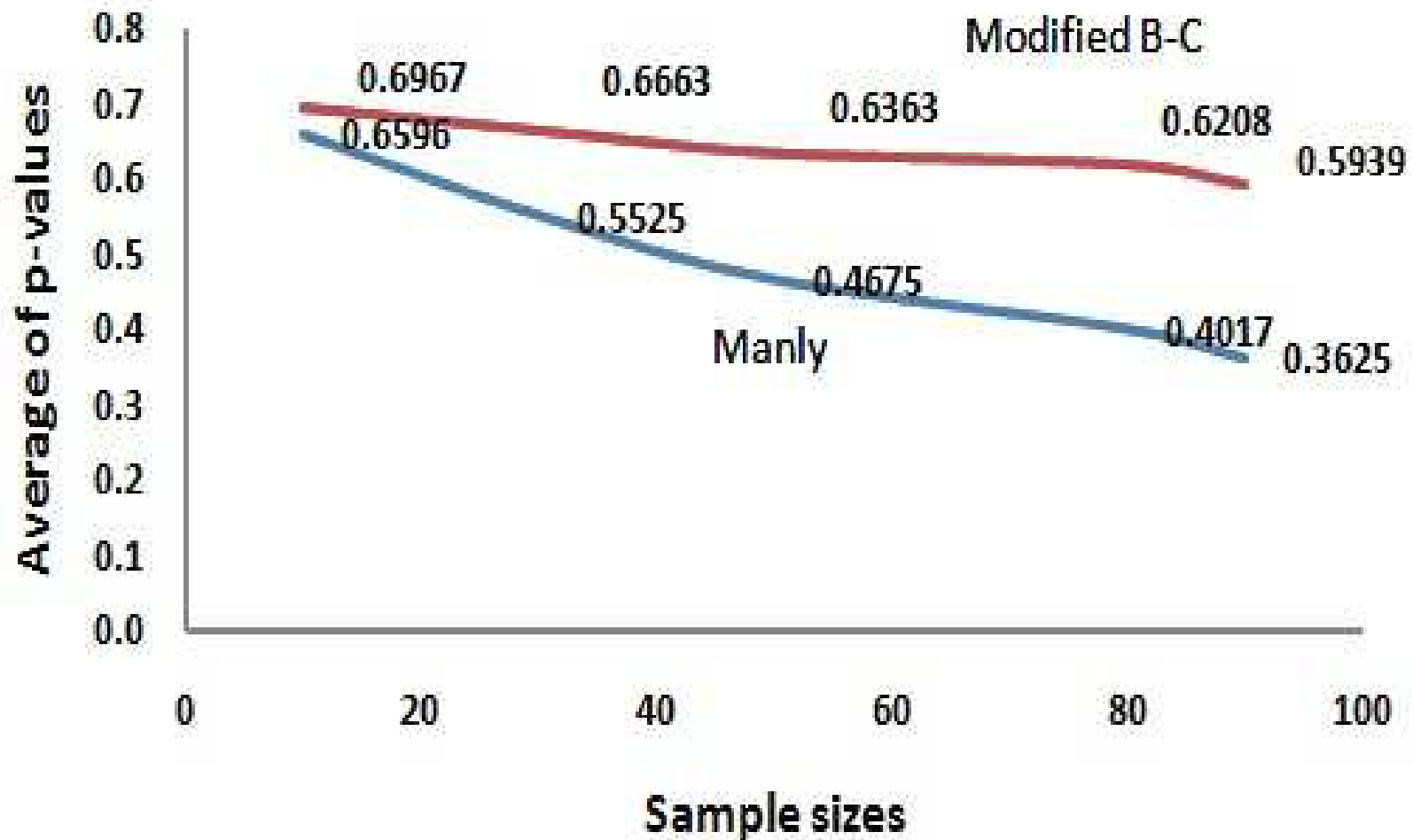
Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=90$	Manly	0.3398	0.3732	0.4562
	Modified	0.3430	0.3921	0.4509
$n_1=5, n_2=10,$	Manly	0.8369	0.7793	0.7589
$n_3=15$	Modified	0.8383	0.7803	0.7566



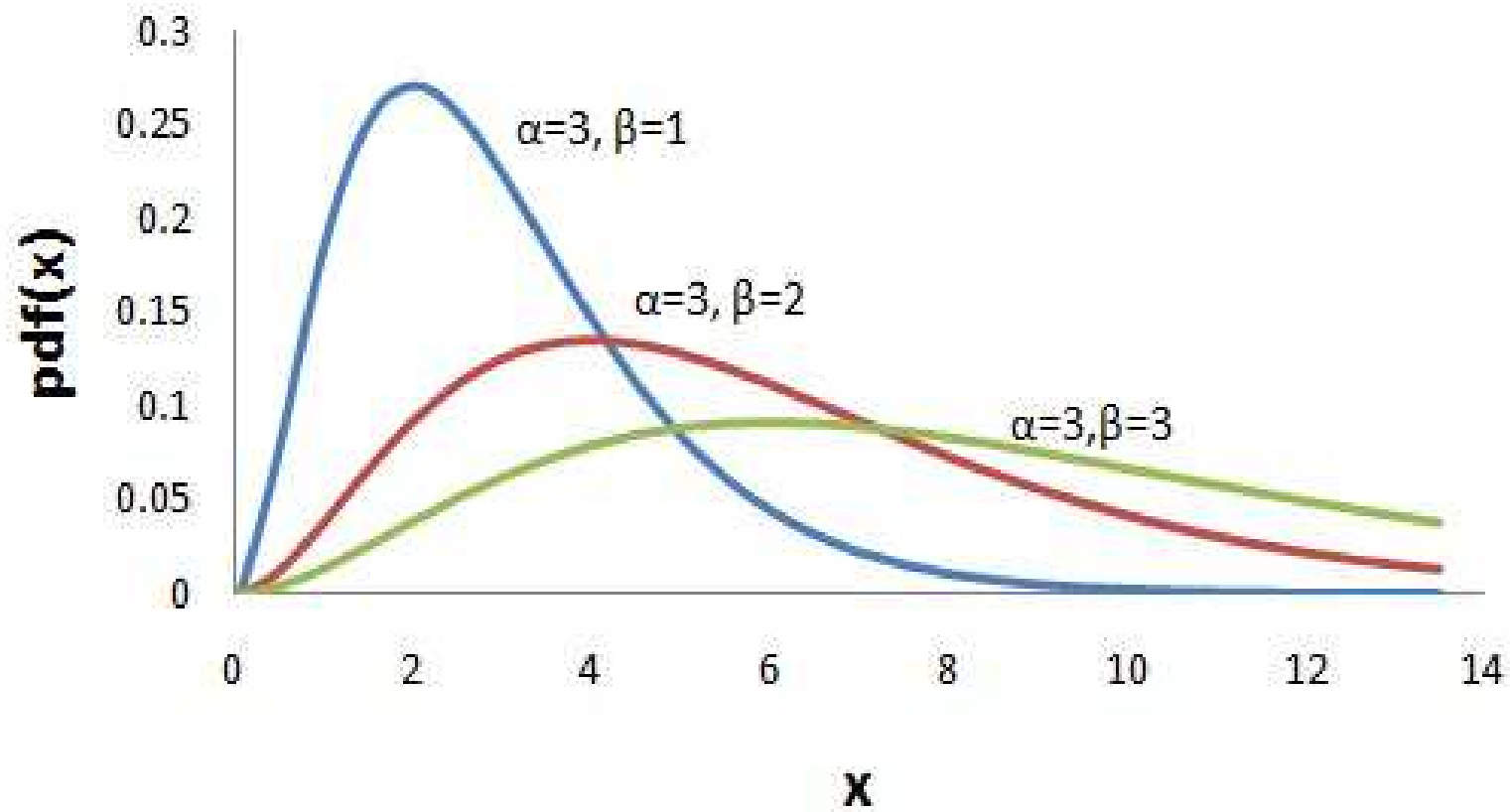
Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_1=5, n_2=15,$	Manly	0.8430	0.7495	0.6106
$n_3=25$	Modified	0.8445	0.7502	0.6114
$n_1=10, n_2=30$	Manly	0.7809	0.6447	0.5748
$n_3=50$	Modified	0.7873	0.6476	0.5731

Average of the p-values for the Levene-test



Graph of Gamma Distribution



Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=10$	Manly	0.7758	0.7764	0.7693
	Modified	0.7803	0.7977	0.7707
$n_i=30$	Manly	0.6358	0.6288	0.6030
	Modified	0.6529	0.6315	0.6085



Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=50$	Manly	0.5022	0.4897	0.4623
	Modified	0.5251	0.4929	0.4739
$n_i=80$	Manly	0.3769	0.3499	0.3208
	Modified	0.4057	0.3524	0.3325



Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=90$	Manly	0.3389	0.3077	0.3961
	Modified	0.3685	0.3105	0.3104
$n_1=5, n_2=10,$	Manly	0.8329	0.7940	0.7447
$n_3=15$	Modified	0.8348	0.7941	0.7455

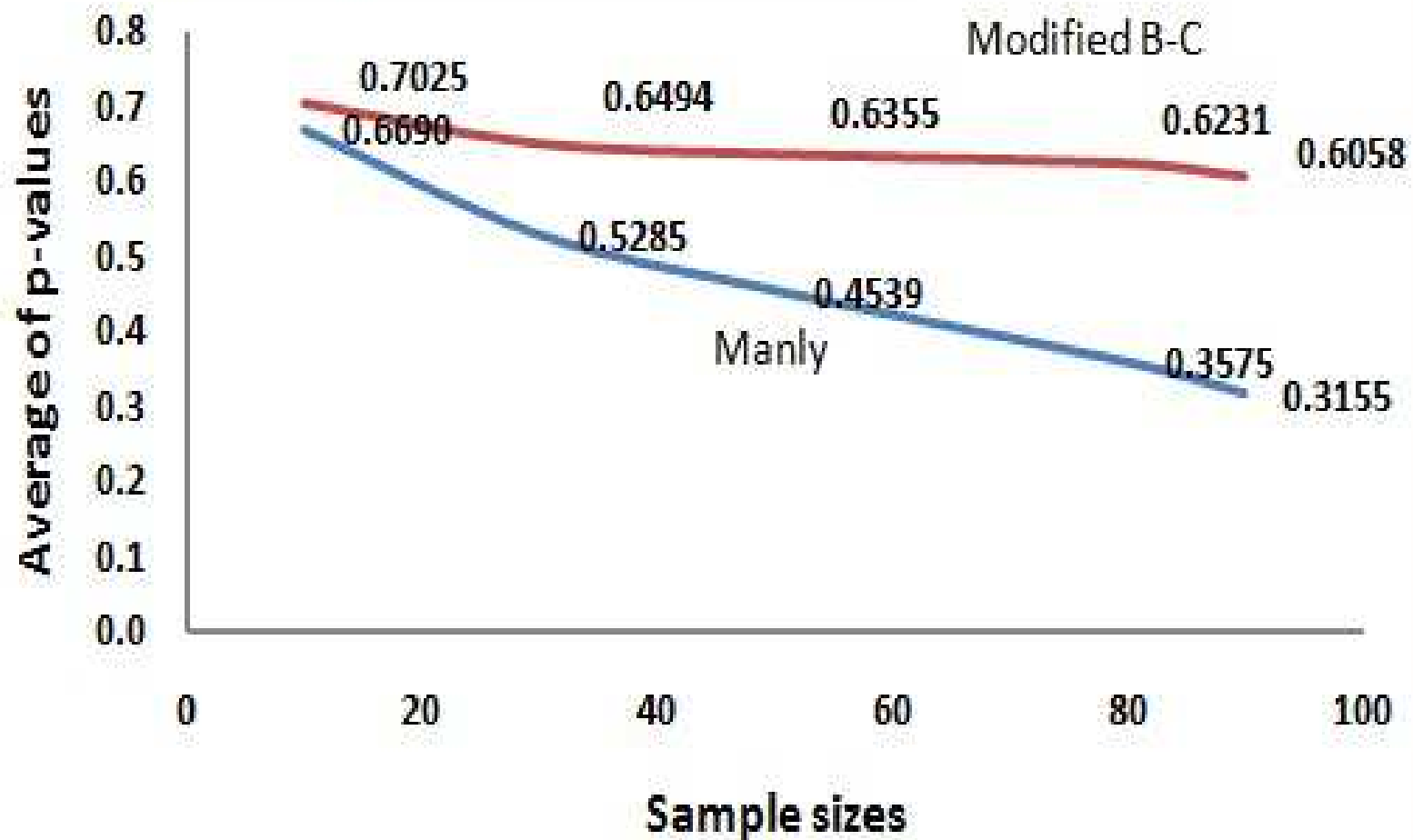


Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_1=5, n_2=15,$	Manly	0.8407	0.7625	0.6740
$n_3=25$	Modified	0.8440	0.7624	0.6782
$n_1=10, n_2=30$	Manly	0.7978	0.6767	0.5257
$n_3=50$	Modified	0.8054	0.6769	0.5279



Average of the p-values for the Levene-test



SIMULATION STUDY FOR THE EXPONENTIAL DATA

- ▶ The possible value for study is set as follows:
 k = number of the Exponential populations = 3,
 n_i = sample size from the i th Exponential population is between 5 and 90,
 β_i = scale parameter of the i th Exponential population is 2 and 9.

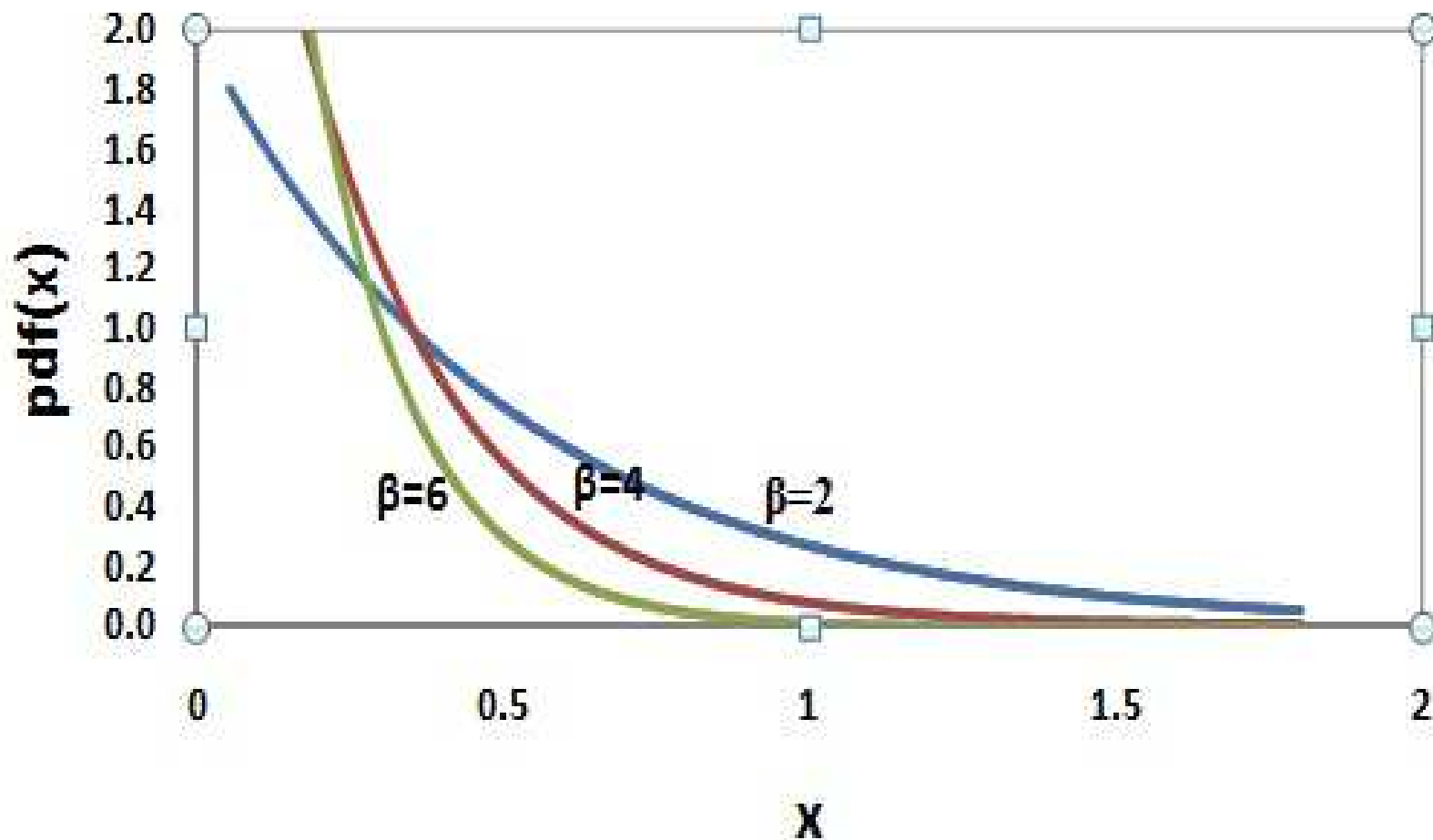


The Results

- ▶ The results of the goodness- of-fit tests and the tests of homogeneity of variances with 1,000 replicated samples of various sizes are as follows



Graph of Exponential Distribution



Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=10$	Manly	0.7083	0.8104	0.8237
	Modified	0.8206	0.8409	0.8381
$n_i=30$	Manly	0.4432	0.7083	0.6802
	Modified	0.7229	0.7586	0.6858



Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=50$	Manly	0.2987	0.6234	0.5810
	Modified	0.6701	0.7056	0.5921
$n_i=80$	Manly	0.1496	0.4843	0.4443
	Modified	0.5457	0.6254	0.4558



Averages of the p-Values for K-S Test

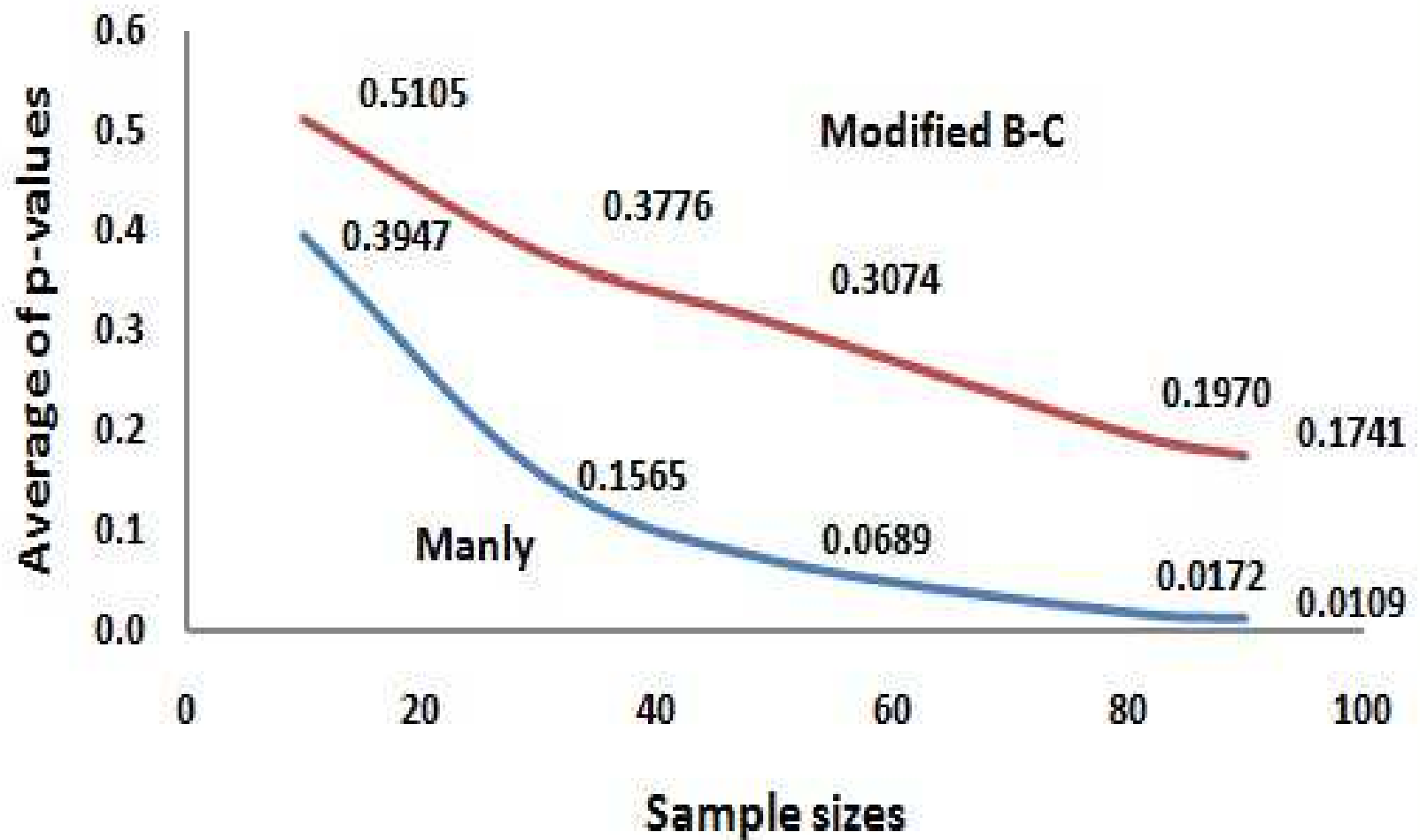
Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=90$	Manly	0.1210	0.4596	0.4160
	Modified	0.5246	0.5970	0.4091
$n_1=5, n_2=10,$	Manly	0.7989	0.7804	0.7669
$n_3=15$	Modified	0.8414	0.8223	0.8028



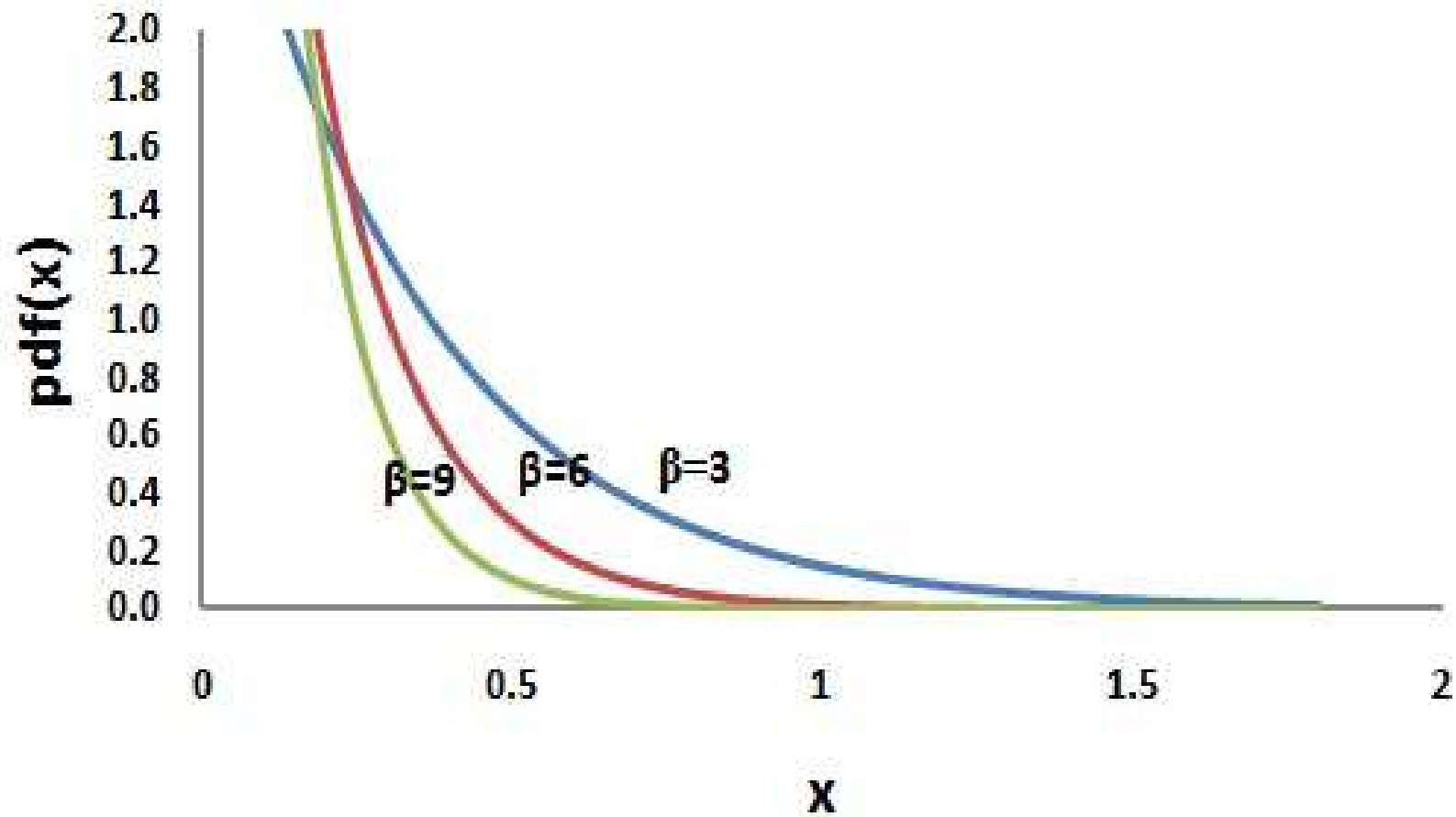
Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_1=5, n_2=15,$	Manly	0.8037	0.7252	0.7136
$n_3=25$	Modified	0.8449	0.8111	0.7880
$n_1=10, n_2=30$	Manly	0.6872	0.5985	0.5346
$n_3=50$	Modified	0.7896	0.7640	0.6814

Average of the p-values for the Levene-test



Graph of Exponential Distribution



Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=10$	Manly	0.7042	0.7907	0.8086
	Modified	0.8049	0.8265	0.8258
$n_i=30$	Manly	0.4407	0.6572	0.6729
	Modified	0.6948	0.7466	0.7239



Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=50$	Manly	0.2975	0.5552	0.5377
	Modified	0.6154	0.6846	0.6189
$n_i=80$	Manly	0.1553	0.4347	0.4130
	Modified	0.4945	0.6139	0.5249



Averages of the p-Values for K-S Test

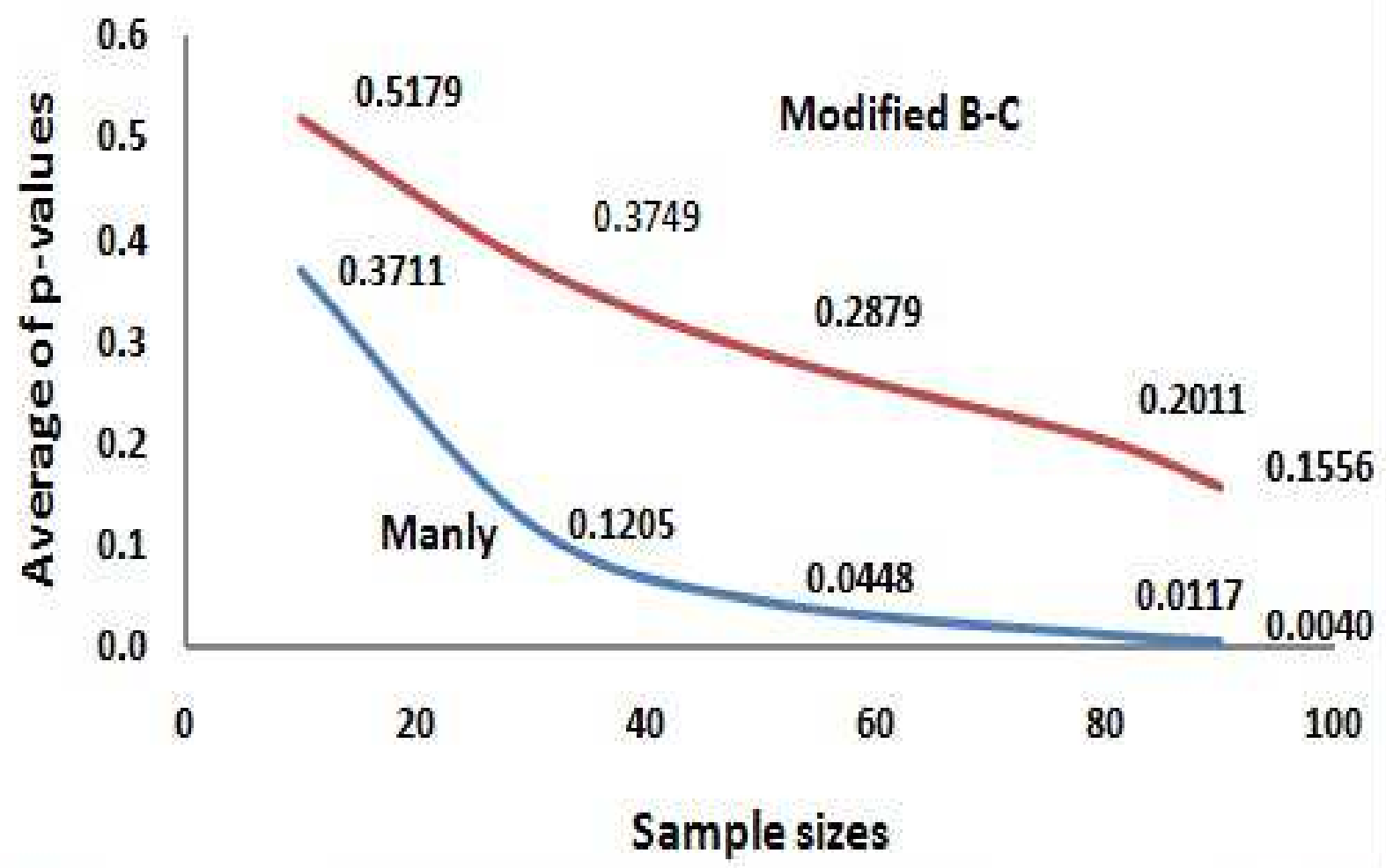
Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_i=90$	Manly	0.1254	0.3935	0.3544
	Modified	0.4706	0.5796	0.4740
$n_1=5, n_2=10,$	Manly	0.8110	0.7851	0.8002
$n_3=15$	Modified	0.8486	0.8213	0.8287



Averages of the p-Values for K-S Test

Sample sizes	Transformations	Averages of the p-Values for K-S Test		
$n_1=5, n_2=15,$	Manly	0.8070	0.7475	0.7508
$n_3=25$	Modified	0.8434	0.8125	0.8033
$n_1=10, n_2=30$	Manly	0.7244	0.6356	0.6413
$n_3=50$	Modified	0.8208	0.7756	0.7265

Average of the p-values for the Levene-test



Conclusions

- ▶ Both two transformations can transform the failure time data to correspond with the basic assumptions.
- ▶ However, It seems that sample sizes affect on Levene test.



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