

Construction of statistic distribution models for nonparametric goodness-of-fit tests in testing composite hypotheses: the computer approach

B.Yu. Lemeshko and S.B. Lemeshko

Abstract— In composite hypotheses testing, when the estimate of the scalar or vector parameter of the probabilities distribution laws is calculated by the same sample, the nonparametric goodness-of-fit Kolmogorov, Cramer-Mises-Smirnov, Anderson-Darling tests lose the free distribution property. In testing of composite hypotheses, the conditional distribution law of the statistic is affected by a number of factors: the form of the observed probabilities distribution law corresponding to the true testable hypothesis; the type of the parameter estimated and the number of parameters to be estimated; sometimes, it is a specific value of the parameter (e.g., in the case of gamma-distribution and beta-distribution families); the method of parameter estimation. In this paper we present more precise results (tables of percentage points and statistic distribution models) for the nonparametric goodness-of-fit tests in testing composite hypotheses using the maximum likelihood estimate (MLE) for some probabilities distribution laws.

Statistic distributions of the nonparametric goodness-of-fit tests are investigated by the methods of statistical simulation. Constructed empirical statistic distributions are approximated with analytical law models.

Index Terms—Goodness-of-fit test, Composite hypotheses testing, Kolmogorov test, Cramer-Mises-Smirnov test, Anderson-Darling test.

I. INTRODUCTION

IN composite hypotheses testing of the form $H_0: F(x) \in F(x, \theta), \theta \in \Theta$, when the estimate $\hat{\theta}$ of the scalar or vector distribution parameter $F(x, \theta)$ is calculated by the same sample, the nonparametric goodness-of-fit Kolmogorov, ω^2 Cramer-Mises-Smirnov, Ω^2 Anderson-Darling tests lose the free distribution property.

The value

Manuscript received April 30, 2008. This work was supported in part by the Russian Foundation for Basic Research, project no. 06-01-00059a.

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$$D_n = \sup_{|x| < \infty} |F_n(x) - F(x, \theta)|,$$

where $F_n(x)$ is the empirical distribution function, n is the sample size, is used in Kolmogorov test as a distance between the empirical and theoretical laws. In testing hypotheses, a statistic with Bolshev [1] correction of the form

$$S_K = \frac{6nD_n + 1}{6\sqrt{n}}, \quad (1)$$

where $D_n = \max(D_n^+, D_n^-)$,

$$D_n^+ = \max_{1 \leq i \leq n} \left\{ \frac{i}{n} - F(x_i, \theta) \right\}, \quad D_n^- = \max_{1 \leq i \leq n} \left\{ F(x_i, \theta) - \frac{i-1}{n} \right\},$$

n is the sample size, x_1, x_2, \dots, x_n are sample values in increasing order is usually used. The distribution of statistic (1) in testing simple hypotheses obeys the Kolmogorov distribution law $K(S) = \sum_{k=-\infty}^{\infty} (-1)^k e^{-2k^2 S^2}$.

In ω^2 Cramer-Mises-Smirnov test, one uses a statistic of the form

$$S_\omega = n\omega_n^2 = \frac{1}{12n} + \sum_{i=1}^n \left\{ F(x_i, \theta) - \frac{2i-1}{2n} \right\}^2, \quad (2)$$

and in test of Ω^2 Anderson-Darling type, the statistic of the form

$$S_\Omega = -n - 2 \sum_{i=1}^n \left\{ \frac{2i-1}{2n} \ln F(x_i, \theta) + \left(1 - \frac{2i-1}{2n} \right) \ln(1 - F(x_i, \theta)) \right\}. \quad (3)$$

In testing a simple hypothesis, statistic (2) obeys the distribution [1] of the form

$$a1(S) = \frac{1}{\sqrt{2s}} \sum_{j=0}^{\infty} \frac{\Gamma(j+1/2)\sqrt{4j+1}}{\Gamma(1/2)\Gamma(j+1)} \exp\left\{-\frac{(4j+1)^2}{16S}\right\} \times \left\{ I_{-\frac{1}{4}}\left[\frac{(4j+1)^2}{16S}\right] - I_{\frac{1}{4}}\left[\frac{(4j+1)^2}{16S}\right] \right\}$$

where $I_{-\frac{1}{4}}(\cdot)$, $I_{\frac{1}{4}}(\cdot)$ - modified Bessel function,

$$I_\nu(z) = \sum_{k=0}^{\infty} \frac{\left(\frac{z}{2}\right)^{\nu+2k}}{\Gamma(k+1)\Gamma(k+\nu+1)}, \quad |z| < \infty, \quad |\arg z| < \pi,$$

and statistic (3) obeys the distribution [1] of the form

$$a_2(S) = \frac{\sqrt{2\pi}}{S} \sum_{j=0}^{\infty} (-1)^j \frac{\Gamma\left(j + \frac{1}{2}\right)(4j+1)}{\Gamma\left(\frac{1}{2}\right)\Gamma(j+1)} \exp\left\{-\frac{(4j+1)^2 \pi^2}{8S}\right\} \times \int_0^{\infty} \exp\left\{\frac{S}{8(y^2+1)} - \frac{(4j+1)^2 \pi^2 y^2}{8S}\right\} dy$$

II. STATISTIC DISTRIBUTIONS OF THE TESTS IN TESTING COMPOSITE HYPOTHESES

In composite hypotheses testing, the conditional distribution law of the statistic $G(S|H_0)$ is affected by a number of factors: the form of the observed law $F(x, \theta)$ corresponding to the true hypothesis H_0 ; the type of the parameter estimated and the number of parameters to be estimated; sometimes, it is a specific value of the parameter (e.g., in the case of gamma-distribution and beta-distribution families); the method of parameter estimation. The distinctions in the limiting distributions of the same statistics in testing simple and composite hypotheses are so significant that we cannot neglect them. For example, Figure 1 shows distributions of the Anderson-Darling statistic (3) while testing the composite hypotheses subject to different laws using maximum likelihood estimates (MLE) of two parameters.

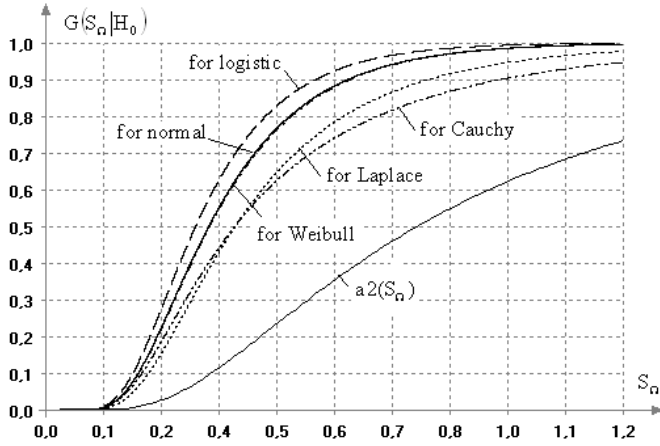


Fig. 1. The Anderson-Darling statistic (3) distributions for testing composite hypotheses with calculating MLE of two law parameters

Figure 2 illustrates the dependence of Kolmogorov test statistic (1) distribution upon the type and the number of estimated parameters by the example of *Su*-Jonson law.

The [2] was a pioneer in investigating statistic distributions of the nonparametric goodness-of-fit tests with composite hypotheses. Then, for the solution to this problem, various approaches were used [3-10].

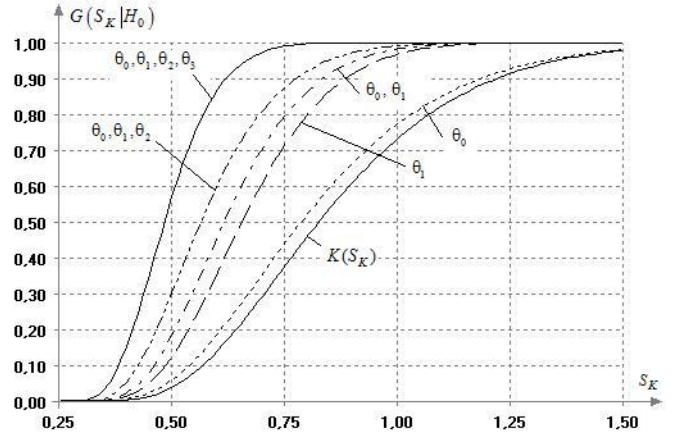


Fig. 2. The Cramer-Mises-Smirnov statistic (2) distributions for testing composite hypotheses with calculating MLE of *Su*-Jonson distribution law parameters

In our research [11-14] statistic distributions of the nonparametric goodness-of-fit tests are investigated by the methods of statistical simulating, and for constructed empirical distributions approximate models of law are found. The results obtained were used to develop recommendations for standardization R 50.1.037-2002 [15].

III. IMPROVEMENT OF STATISTIC DISTRIBUTION MODELS OF THE NONPARAMETRIC GOODNESS-OF-FIT TESTS

In this paper we present more precise results (tables of percentage points and statistic distribution models) for the nonparametric goodness-of-fit tests in testing composite hypotheses using the maximum likelihood estimate (MLE).

These investigations have been based on the developed software support. The software support allows of investigating probability regularities with the Monte-Carlo methods and constructing approximate analytical models for these regularities.

Table 1 contains a list of distributions relative to which we can test composite fit hypotheses using the constructed approximations of the limiting statistic distributions.

The tables of percentage points and statistic distributions models were constructed by modeled statistic samples with the size $N = 10^6$. In this case, the samples of pseudorandom variables, belonging to $F(x, \theta)$, were generated with the size $n = 10^3$.

Distributions $G(S|H_0)$ of the Kolmogorov statistic are best approximated by gamma-distributions family $\gamma(\theta_0, \theta_1, \theta_2)$ (see Table 1) or by the family of the III type beta-distributions with the density function

$$B_3 \theta_0, \theta_1, \theta_2, \theta_3, \theta_4 = \frac{\theta_2^{\theta_0}}{\theta_3 B \theta_0, \theta_1} \frac{\left(\frac{x-\theta_4}{\theta_3}\right)^{\theta_0-1} \left(1 - \frac{x-\theta_4}{\theta_3}\right)^{\theta_1-1}}{\left[1 + \theta_2 - 1 \frac{x-\theta_4}{\theta_3}\right]^{\theta_0+\theta_1}}$$

Table 1. Random variable distribution

Random variable distribution	Density function $f(x, \theta)$
Exponential	$\frac{1}{\theta_0} e^{-x/\theta_0}$
Seminormal	$\frac{2}{\theta_0 \sqrt{2\pi}} e^{-x^2/2\theta_0^2}$
Rayleigh	$\frac{x}{\theta_0^2} e^{-x^2/2\theta_0^2}$
Maxwell	$\frac{2x^2}{\theta_0^3 \sqrt{2\pi}} e^{-x^2/2\theta_0^2}$
Laplace	$\frac{1}{2\theta_0} e^{- x-\theta_1 /\theta_0}$
Normal	$\frac{1}{\theta_0 \sqrt{2\pi}} e^{-\frac{(x-\theta_1)^2}{2\theta_0^2}}$
Log-normal	$\frac{1}{x\theta_0 \sqrt{2\pi}} e^{-(\ln x - \theta_1)^2/2\theta_0^2}$
Cauchy	$\frac{\theta_0}{\pi[\theta_0^2 + (x - \theta_1)^2]}$
Logistic	$\frac{\pi}{\theta_0 \sqrt{3}} \exp\left\{-\frac{\pi(x - \theta_1)}{\theta_0 \sqrt{3}}\right\} / \left[1 + \exp\left\{-\frac{\pi(x - \theta_1)}{\theta_0 \sqrt{3}}\right\}\right]^2$
Extreme-value (maximum)	$\frac{1}{\theta_0} \exp\left\{-\frac{x - \theta_1}{\theta_0} - \exp\left(-\frac{x - \theta_1}{\theta_0}\right)\right\}$
Extreme-value (minimum)	$\frac{1}{\theta_0} \exp\left\{\frac{x - \theta_1}{\theta_0} - \exp\left(\frac{x - \theta_1}{\theta_0}\right)\right\}$
Weibull	$\frac{\theta_0 x^{\theta_0 - 1}}{\theta_1^{\theta_0}} \exp\left\{-\left(\frac{x}{\theta_1}\right)^{\theta_0}\right\}$
<i>Sb</i> -Johnson <i>Sb</i> $\theta_0, \theta_1, \theta_2, \theta_3$	$\frac{\theta_1 \theta_2}{(x - \theta_3)(\theta_2 + \theta_3 - x)} \times \exp\left\{-\frac{1}{2}\left[\theta_0 - \theta_1 \ln \frac{x - \theta_3}{\theta_2 + \theta_3 - x}\right]^2\right\}$
<i>Sl</i> -Johnson <i>Sl</i> $\theta_0, \theta_1, \theta_2, \theta_3$	$\frac{\theta_1}{x - \theta_3} \exp\left\{-\frac{1}{2}\left[\theta_0 + \theta_1 \ln \frac{x - \theta_3}{\theta_2}\right]^2\right\}$
<i>Su</i> -Johnson <i>Su</i> $\theta_0, \theta_1, \theta_2, \theta_3$	$\frac{\theta_1}{\sqrt{2\pi} \sqrt{x - \theta_3}^2 + \theta_2^2} \times \exp\left\{-\frac{1}{2}\left[\theta_0 + \theta_1 \ln \left[\frac{x - \theta_3}{\theta_2} + \sqrt{\left(\frac{x - \theta_3}{\theta_2}\right)^2 + 1}\right]\right]^2\right\}$

Gamma-distribution $\gamma \theta_0, \theta_1, \theta_2$	$\frac{1}{\theta_1^{\theta_0} \Gamma(\theta_0)} x - \theta_2^{\theta_0 - 1} e^{-x - \theta_2 / \theta_1}$
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And distributions of the Cramer-Mises-Smirnov and the Anderson-Darling statistics are well approximated by the family of the *Sb*-Johnson distributions *Sb* $\theta_0, \theta_1, \theta_2, \theta_3$ (see Table 1) or by the family of the III type beta-distributions *B*₃ $\theta_0, \theta_1, \theta_2, \theta_3, \theta_4, \dots$

Upper percentage points and constructed models of the limiting statistic distributions of the Kolmogorov test when MLE are used are presented in Table 2 [for law: exponential, seminormal, Rayleigh, Maxwell, Laplace, normal, log-normal, Cauchy, logistic, extreme-value (maximum and minimum), Weibull].

Table 2. Upper percentage points and models of limiting statistic distributions of the Kolmogorov's test when MLE are used

Random variable distribution	Parameter estimated	Percentage points			Model
		0.1	0.05	0.01	
Exponential & Rayleigh	Scale	0.995	1.094	1.292	$\gamma(5.1092; 0.0861; 0.2950)$
Seminormal	Scale	1.051	1.160	1.381	$\gamma(4.5462; 0.1001; 0.3100)$
Maxwell	Scale	0.969	1.062	1.251	$\gamma(5.4566; 0.0794; 0.2870)$
Laplace	Scale	1.177	1.313	1.586	$\gamma(3.3950; 0.1426; 0.3405)$
	Shift	0.957	1.044	1.223	$\gamma(5.1092; 0.0861; 0.2950)$
	2 param-s	0.863	0.940	1.096	$\gamma(6.2949; 0.0624; 0.2613)$
Normal & Log-normal	Scale	1.191	1.327	1.600	$\gamma(3.5609; 0.1401; 0.3375)$
	Shift	0.888	0.963	1.114	$\gamma(7.5304; 0.0580; 0.2400)$
	2 param-s	0.835	0.909	1.057	$\gamma(6.4721; 0.0580; 0.2620)$
Cauchy	Scale	1.137	1.275	1.550	$\gamma(3.0987; 0.1463; 0.3350)$
	Shift	0.975	1.070	1.260	$\gamma(5.9860; 0.0780; 0.2528)$
	2 param-s	0.815	0.893	1.048	$\gamma(5.3642; 0.0654; 0.2600)$
Logistic	Scale	1.180	1.316	1.589	$\gamma(3.4954; 0.1411; 0.3325)$
	Shift	0.837	0.907	1.046	$\gamma(7.6325; 0.0531; 0.2368)$
	2 param-s	0.747	0.805	0.923	$\gamma(7.5402; 0.0451; 0.2422)$
Extreme-value & Weibull	Scale ¹⁾	1.182	1.316	1.583	$\gamma(3.6805; 0.1355; 0.3350)$
	Shift ²⁾	0.995	1.093	1.292	$\gamma(5.2194; 0.0848; 0.2920)$ ¹⁾
	2 param-s	0.824	0.895	1.037	$\gamma(6.6012; 0.0563; 0.2598)$

Note. ¹⁾ - we estimated the Weibull distribution form parameter, ²⁾ - the Weibull distribution scale parameter.

For the same distribution laws upper percentage points and constructed models of the limiting statistic distributions of the Cramer-von Mises-Smirnov test are presented in Table 3, for Anderson-Darling test - in Table 4.

In Table 5 there are upper percentage points and models of limiting statistic distributions of nonparametric goodness-of-fit when MLE are used in the case of *Sb*-Johnson distribution, in Table 6 – in the case of *Sl*-Johnson distribution, in Table 7 – in the case of *Su*-Johnson distribution.

Table 3. Upper percentage points and models of limiting statistic distributions of the Cramer-Mises-Smirnov’s test when MLE are used

Random variable distribution	Parameter estimated	Percentage points			Model
		0.1	0.05	0.01	
Exponential & Rayleigh	Scale	0.174	0.221	0.337	$Sb(3.3738;1.2145;1.0792;0.011)$
Seminormal	Scale	0.205	0.266	0.415	$Sb(3.527;1.1515;1.5527;0.012)$
Maxwell	Scale	0.162	0.204	0.306	$Sb(3.353;1.220;0.9786;0.0118)$
Laplace	Scale	0.323	0.438	0.719	$Sb(3.2262;0.9416;2.703;0.015)$
	Shift	0.151	0.187	0.267	$Sb(2.9669;1.2534;0.6936;0.01)$
	2 param-s	0.115	0.144	0.214	$Sb(3.768;1.2865;0.8336;0.0113)$
Normal & Log-normal	Scale	0.327	0.443	0.727	$Sb(3.153;0.9448;2.5477;0.016)$
	Shift	0.134	0.165	0.238	$Sb(3.243;1.315;0.6826;0.0095)$
	2 param-s	0.103	0.126	0.178	$Sb(4.3950;1.4428;0.915;0.009)$
Cauchy	Scale	0.316	0.430	0.711	$Sb(3.1895;0.9134;2.690;0.013)$
	Shift	0.172	0.216	0.319	$Sb(2.359;1.0732;0.595;0.0129)$
	2 param-s	0.129	0.170	0.271	$Sb(3.4364;1.0678;1.000;0.011)$
Logistic	Scale	0.323	0.438	0.719	$Sb(3.264;0.9581;2.7046;0.014)$
	Shift	0.119	0.148	0.216	$Sb(4.0026;1.2853;1.00;0.0122)$
	2 param-s	0.081	0.098	0.135	$Sb(3.2137;1.3612;0.36;0.0105)$
Extreme-value & Weibull	Scale ¹⁾	0.320	0.431	0.704	$Sb(3.343;0.9817;2.753;0.015)^1$
	Shift ²⁾	0.174	0.221	0.336	$Sb(3.498;1.2236;1.1632;0.01)^2$
	2 param-s	0.102	0.124	0.174	$Sb(3.3854;1.4453;0.4986;0.007)$

Note. ¹⁾ - we estimated the Weibull distribution form parameter, ²⁾ - the Weibull distribution scale parameter.

Table 4. Upper percentage points and models of limiting statistic distributions of the Anderson-Darling’s test when MLE are used

Random variable distribution	Parameter estimated	Percentage points			Model
		0.1	0.05	0.01	
Exponential & Rayleigh	Scale	1.060	1.319	1.954	$Sb(3.8386;1.3429;7.500;0.090)$
Seminormal	Scale	1.188	1.499	2.267	$Sb(4.2019;1.2918;11.500;0.100)$
Maxwell	Scale	1.010	1.247	1.832	$Sb(3.9591;1.3296;7.800;0.101)$
Laplace	Scale	1.726	2.286	3.684	$Sb(4.3260;1.0982;27.00;0.110)$
	Shift	1.070	1.301	1.832	$Sb(3.1506;1.3352;4.9573;0.096)$
	2 param-s	0.797	0.982	1.440	$Sb(3.8071;1.3531;5.1809;0.10)$
Normal & Log-normal	Scale	1.745	2.309	3.706	$Sb(4.3271;1.0895;28.000;0.12)$
	Shift	0.892	1.087	1.551	$Sb(3.3085;1.4043;4.2537;0.08)$
	2 param-s	0.629	0.750	1.030	$Sb(3.5601;1.4846;3.0987;0.08)$
Cauchy	Scale	1.716	2.277	3.673	$Sb(3.7830;1.0678;18.0;0.11)$
	Shift	1.215	1.512	2.211	$Sb(3.4814;1.2375;7.810;0.1)$
	2 param-s	0.948	1.226	1.913	$Sb(3.290;1.129;5.837;0.099)$
Logistic	Scale	1.724	2.285	3.682	$Sb(3.516;1.054;14.748;0.117)$
	Shift	0.856	1.043	1.495	$Sb(5.1316;1.5681;10.0;0.065)$
	2 param-s	0.562	0.665	0.903	$Sb(3.409;1.434;2.448;0.095)$
Extreme-value & Weibull	Scale ¹⁾	1.723	2.273	3.634	$Sb(3.512;1.064;14.496;0.125)^1$
	Shift ²⁾	1.059	1.318	1.952	$Sb(4.799;1.402;13.0;0.085)^2$
	2 param-s	0.634	0.755	1.040	$Sb(3.4830;1.5138;3.00;0.07)$

Note. ¹⁾ - we estimated the Weibull distribution form parameter, ²⁾ - the Weibull distribution scale parameter.

Table 5. Upper percentage points and models of limiting statistic distributions of nonparametric goodness-of-fit when MLE are used in the case of *Sb*-Johnson distribution

Parameter estimated	Percentage points			Model
	0.9	0.95	0.99	
for Kolmogorov’s test				
θ_0	0.888	0.963	1.115	$B_3(6.3484, 7.4913, 2.3663, 1.4790, 0.27)$
θ_1	1.189	1.326	1.600	$B_3(6.8242, 4.7737, 5.2621, 2.3878, 0.27)$
θ_0 & θ_1	0.836	0.909	1.058	$B_3(6.6559, 8.1766, 2.9405, 1.6143, 0.27)$
for Cramer-Mises-Smirnov’s test				
θ_0	0.134	0.165	0.238	$B_3(4.2304, 3.806, 13.1934, 0.6908, 0.0086)$
θ_1	0.327	0.442	0.724	$B_3(2.9153, 2.0048, 33.416, 2.0782, 0.0114)$
θ_0 & θ_1	0.104	0.126	0.179	$B_3(4.3897, 4.0574, 12.101, 0.5119, 0.0086)$
for Anderson-Darling’s test				
θ_0	0.893	1.087	1.553	$B_3(4.2657, 4.3788, 11.4946, 4.655, 0.084)$
θ_1	1.741	2.309	3.702	$B_3(4.1703, 2.3363, 42.083, 12.602, 0.088)$
θ_0 & θ_1	0.631	0.751	1.034	$B_3(4.0891, 5.9708, 9.6497, 4.0000, 0.082)$

Table 6. Upper percentage points and models of limiting statistic distributions of nonparametric goodness-of-fit when MLE are used in the case of *Sl*-Johnson distribution

Parameter estimated	Percentage points			Model
	0.9	0.95	0.99	
for Kolmogorov’s test				
θ_0	0.888	0.963	1.115	$B_3(6.3484, 7.4913, 2.3663, 1.4790, 0.27)$
θ_1	1.189	1.326	1.600	$B_3(6.8242, 4.7737, 5.2621, 2.3878, 0.27)$
θ_2	0.888	0.963	1.115	$B_3(6.3484, 7.4913, 2.3663, 1.4790, 0.27)$
θ_0, θ_1	0.836	0.909	1.058	$B_3(6.6559, 8.1766, 2.9405, 1.6143, 0.27)$
θ_0, θ_2	0.888	0.963	1.115	$B_3(6.3484, 7.4913, 2.3663, 1.4790, 0.27)$
θ_1, θ_2	0.836	0.909	1.058	$B_3(6.6559, 8.1766, 2.9405, 1.6143, 0.27)$
$\theta_0, \theta_1, \theta_2$	0.836	0.909	1.058	$B_3(6.6559, 8.1766, 2.9405, 1.6143, 0.27)$
for Cramer-Mises-Smirnov’s test				
θ_0	0.134	0.165	0.238	$B_3(4.2304, 3.8058, 13.193, 0.6908, 0.0086)$
θ_1	0.327	0.442	0.724	$B_3(2.9153, 2.0048, 33.414, 2.0782, 0.0114)$
θ_2	0.134	0.165	0.238	$B_3(4.2304, 3.8058, 13.193, 0.6908, 0.0086)$
θ_0, θ_1	0.104	0.126	0.179	$B_3(4.3897, 4.0574, 12.101, 0.5119, 0.0086)$
θ_0, θ_2	0.134	0.165	0.238	$B_3(4.2304, 3.8058, 13.193, 0.6908, 0.0086)$
θ_1, θ_2	0.104	0.126	0.179	$B_3(4.3897, 4.0574, 12.101, 0.5119, 0.0086)$
$\theta_0, \theta_1, \theta_2$	0.104	0.126	0.179	$B_3(4.3897, 4.0574, 12.101, 0.5119, 0.0086)$
for Anderson-Darling’s test				
θ_0	0.893	1.087	1.553	$B_3(4.2657, 4.3788, 11.4946, 4.6551, 0.084)$
θ_1	1.741	2.309	3.702	$B_3(4.1703, 2.3363, 42.083, 12.6019, 0.088)$
θ_2	0.893	1.087	1.553	$B_3(4.2657, 4.3788, 11.4946, 4.6551, 0.084)$
θ_0, θ_1	0.631	0.751	1.034	$B_3(4.0891, 5.9708, 9.6497, 4.0000, 0.082)$
θ_0, θ_2	0.893	1.087	1.553	$B_3(4.2657, 4.3788, 11.4946, 4.6551, 0.084)$
θ_1, θ_2	0.631	0.751	1.034	$B_3(4.0891, 5.9708, 9.6497, 4.0000, 0.082)$
$\theta_0, \theta_1, \theta_2$	0.631	0.751	1.034	$B_3(4.0891, 5.9708, 9.6497, 4.0000, 0.082)$

Table 7. Upper percentage points and models of limiting statistic distributions of nonparametric goodness-of-fit when MLE are used in the case of *Su*-Johnson distribution

Parameter estimated	Percentage points			Model
	0.9	0.95	0.99	
θ_0	0.888	0.963	1.115	$B_3(6.3484,7.4913,2.3663,1.4790,0.27)$
θ_1	1.189	1.326	1.600	$B_3(6.8242,4.7737,5.2621,2.3878,0.27)$
θ_2	1.161	1.300	1.576	$B_3(5.3417,4.6440,4.7448,2.3802,0.29)$
θ_3	0.880	0.960	1.122	$B_3(6.6252,7.4025,3.0590,1.6516,0.27)$
θ_0, θ_1	0.836	0.909	1.058	$B_3(6.4792,7.0243,2.8437,1.4260,0.27)$
θ_0, θ_2	0.798	0.872	1.024	$B_3(6.4496,6.7714,3.3119,1.4226,0.27)$
θ_0, θ_3	0.802	0.875	1.023	$B_3(6.3069,6.1065,3.2916,1.3317,0.27)$
θ_1, θ_3	1.142	1.282	1.561	$B_3(5.9751,4.4559,5.6810,2.4123,0.27)$
θ_1, θ_3	0.792	0.858	0.994	$B_3(6.4839,7.0152,2.7376,1.2838,0.27)$
θ_2, θ_3	0.733	0.791	0.910	$B_3(6.2438,6.9161,2.5011,1.0904,0.27)$
$\theta_0, \theta_1, \theta_2$	0.776	0.851	1.007	$B_3(6.2414,6.4027,3.7458,1.4361,0.27)$
$\theta_0, \theta_1, \theta_3$	0.720	0.780	0.901	$B_3(6.4262,6.9732,2.7325,1.1317,0.26)$
$\theta_0, \theta_2, \theta_3$	0.658	0.706	0.806	$B_3(6.1239,7.9516,2.24033,0.9839,0.26)$
$\theta_1, \theta_2, \theta_3$	0.704	0.760	0.878	$B_3(7.1354,8.0363,2.7466,1.1766,0.25)$
$\theta_0, \theta_1, \theta_2, \theta_3$	0.622	0.666	0.755	$B_3(6.6889,8.1712,2.3857,0.9291,0.25)$
θ_0	0.134	0.165	0.238	$B_3(3.6736,3.9355,11.2146,0.6908,0.01)$
θ_1	0.327	0.442	0.724	$B_3(2.9153,2.0048,33.414,2.078,0.0114)$
θ_2	0.318	0.433	0.716	$B_3(2.2077,1.7250,28.4959,1.75,0.015)$
θ_3	0.125	0.154	0.225	$B_3(3.6990,3.8775,11.9942,0.6601,0.01)$
θ_0, θ_1	0.104	0.126	0.179	$B_3(4.3897,4.0574,12.101,0.512,0.0086)$
θ_0, θ_2	0.090	0.110	0.161	$B_3(5.203,3.9325,15.697,0.4659,0.0075)$
θ_0, θ_3	0.104	0.133	0.203	$B_3(5.954,3.1023,30.6943,0.638,0.0071)$
θ_1, θ_2	0.314	0.428	0.711	$B_3(2.4905,1.6985,45.967,2.3084,0.012)$
θ_1, θ_3	0.094	0.113	0.158	$B_3(4.6011,5.7370,19.1580,1.0,0.0075)$
θ_2, θ_3	0.080	0.096	0.137	$B_3(4.7686,4.6085,11.142,0.393,0.0075)$
$\theta_0, \theta_1, \theta_2$	0.083	0.104	0.155	$B_3(5.2574,3.644,19.921,0.4707,0.0075)$
$\theta_0, \theta_1, \theta_3$	0.071	0.086	0.122	$B_3(5.775,4.7935,18.118,0.4777,0.0065)$
$\theta_0, \theta_2, \theta_3$	0.056	0.066	0.089	$B_3(7.350,5.4726,13.745,0.2883,0.0052)$
$\theta_1, \theta_2, \theta_3$	0.073	0.089	0.130	$B_3(5.6379,4.0985,18.5518,0.421,0.007)$
$\theta_0, \theta_1, \theta_2, \theta_3$	0.048	0.056	0.075	$B_3(6.9739,6.6406,13.743,0.315,0.0052)$
θ_0	0.893	1.087	1.553	$B_3(4.2329,4.5369,10.8807,4.655,0.082)$
θ_1	1.741	2.309	3.702	$B_3(4.1703,2.3363,42.083,12.603,0.088)$
θ_2	1.707	2.275	3.667	$B_3(2.6348,1.9774,21.3842,7.75,0.125)$
θ_3	0.952	1.161	1.649	$B_3(3.5597,4.9656,11.4180,6.520,0.092)$
θ_0, θ_1	0.631	0.751	1.034	$B_3(4.0891,5.9708,9.6497,4.0000,0.082)$
θ_0, θ_2	0.577	0.689	0.961	$B_3(5.5368,4.9114,13.1278,3.0625,0.07)$
θ_0, θ_3	0.737	0.920	1.386	$B_3(5.6629,3.4912,25.1600,4.5052,0.07)$
θ_1, θ_2	1.666	2.232	3.627	$B_3(3.8896,1.6253,31.1820,5.80,0.09)$
θ_1, θ_3	0.694	0.842	1.200	$B_3(4.6199,5.2874,19.2708,6.561,0.074)$

θ_2, θ_3	0.642	0.935	1.140	$B_3(4.4276,4.30288,14.669,3.7865,0.08)$
$\theta_0, \theta_1, \theta_2$	0.518	0.627	0.898	$B_3(5.5158,4.3512,14.770,2.6199,0.067)$
$\theta_0, \theta_1, \theta_3$	0.454	0.536	0.733	$B_3(5.3306,5.8858,10.751,2.5087,0.065)$
$\theta_0, \theta_2, \theta_3$	0.396	0.459	0.606	$B_3(5.7098,6.8325,7.9837,1.8803,0.06)$
$\theta_1, \theta_2, \theta_3$	0.585	0.729	1.087	$B_3(5.1840,3.2993,19.364,2.7865,0.073)$
$\theta_0, \theta_1, \theta_2, \theta_3$	0.329	0.378	0.489	$B_3(7.1015,5.8708,7.1323,1.0517,0.05)$

IV. IMPROVEMENT OF STATISTIC DISTRIBUTION MODELS OF THE NONPARAMETRIC GOODNESS-OF-FIT TESTS IN THE CASE OF GAMMA-DISTRIBUTION

In composite hypotheses testing subject to gamma-distribution with the density function

$$f(x, \theta) = \frac{x^{\theta_0-1}}{\theta_1^{\theta_0} \Gamma(\theta_0)} \exp\left(-\frac{x}{\theta_1}\right)$$

limiting statistics distributions of the nonparametric goodness-of-fit tests depend on values of the form parameter θ_0 . For example, Figure 3 illustrates dependence of the Kolmogorov statistic distribution upon the value θ_0 in testing a composite hypothesis in the case of calculating MLE for the scale parameter of gamma-distribution only.

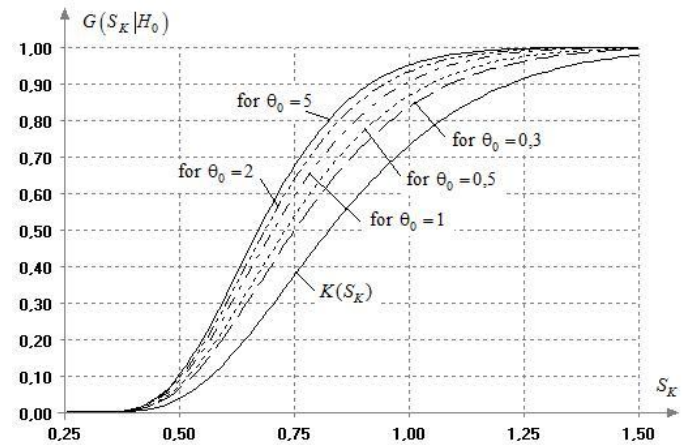


Fig. 3. The Kolmogorov statistic (1) distributions for testing composite hypotheses with calculating MLE of scale parameter depend on the form parameter value of gamma-distribution

Upper percentage points constructed as a result of statistical modeling and constructed models of limiting statistic distributions of the Kolmogorov test when MLE are used in the case of gamma-distribution are given in Table 8, for Cramer-von Mises-Smirnov test – in Table 9, for Anderson-Darling test - in Table 10. In this case statistic distributions are well approximated by the family of the III type beta-distributions $B_3 \theta_0, \theta_1, \theta_2, \theta_3, \theta_4, \dots$

The upper percentage points and the models of statistic distributions, presented at Tables 8-10, improve the results given in recommendations [15].

Table 8. Upper percentage points and models of limiting statistic distributions of the Kolmogorov’s test when MLE are used in the case of gamma-distribution

Value of the form parameter	Parameter estimated	Percentage points			Model
		0.1	0.05	0.01	
0.3	Scale	1.096	1.211	1.444	$B_3(6.687;4.837;4.4047;1.944;0.281)$
	form	0.976	1.070	1.262	$B_3(6.4536;5.7519;3.310;1.650;0.28)$
	2 param-s	0.905	0.990	1.162	$B_3(6.9705;5.678;3.6297;1.507;0.27)$
0.5	Scale	1.051	1.160	1.379	$B_3(6.9356;5.008;4.3582;1.847;0.28)$
	form	0.961	1.052	1.236	$B_3(6.386;5.9685;3.123;1.6154;0.28)$
	2 param-s	0.884	0.965	1.131	$B_3(6.408;5.934;3.206;1.448;0.2774)$
1.0	Scale	0.994	1.095	1.299	$B_3(6.719;5.374;3.7755;1.688;0.282)$
	form	0.936	1.022	1.191	$B_3(6.1176;6.4704;2.6933;1.55;0.28)$
	2 param-s	0.862	0.940	1.097	$B_3(5.603;6.129;2.7065;1.361;0.29)$
2.0	Scale	0.952	1.044	1.228	$B_3(5.8359;22.6032;2.192;4.0;0.282)$
	form	0.915	0.995	1.155	$B_3(6.1387;6.5644;2.602;1.484;0.28)$
	2 param-s	0.849	0.924	1.077	$B_3(5.832;6.145;2.755;1.328;0.2862)$
3.0	Scale	0.933	1.020	1.200	$B_3(5.9055;24.431;2.0996;4.0;0.282)$
	form	0.906	0.985	1.140	$B_3(6.1221;6.613;2.5536;1.459;0.28)$
	2 param-s	0.845	0.919	1.070	$B_3(6.039;6.1276;2.831;1.32;0.2837)$
4.0	Scale	0.923	1.008	1.181	$B_3(5.9419;27.1264;1.915;4.000;282)$
	form	0.901	0.980	1.132	$B_3(6.083;6.7095;2.496;1.4494;0.28)$
	2 param-s	0.843	0.916	1.066	$B_3(6.1584;6.119;2.875;1.317;0.281)$
5.0	Scale	0.917	1.000	1.170	$B_3(5.8774;30.0692;1.720;4.0;0.282)$
	form	0.899	0.977	1.127	$B_3(6.089;6.7265;2.4894;1.443;0.28)$
	2 param-s	0.842	0.915	1.063	$B_3(6.196;6.1114;2.8894;1.314;0.28)$

Table 10. Upper percentage points and models of limiting statistic distributions of the Anderson- Darling’s test when MLE are used in the case of gamma-distribution

Value of the form parameter	Parameter estimated	Percentage points			Model
		0.1	0.05	0.01	
0.3	Scale	1.300	1.655	2.543	$B_3(3.385;2.883;14.684;6.042;0.109)$
	form	1.021	1.258	1.865	$B_3(3.1073;3.704;8.672;4.344;0.112)$
	2 param-s	0.718	0.870	1.233	$B_3(4.532;4.060;10.072;2.921;0.078)$
0.5	Scale	1.183	1.490	2.260	$B_3(5.005;2.936;18.852;5.244;0.077)$
	form	0.993	1.221	1.791	$B_3(3.1104;3.729;8.068;4.013;0.112)$
	2 param-s	0.684	0.824	1.145	$B_3(5.008;4.056;10.029;2.587;0.073)$
1.0	Scale	1.058	1.313	1.955	$B_3(5.031;3.185;15.463;4.380;0.077)$
	form	0.952	1.166	1.696	$B_3(3.115;3.792;7.4813;3.677;0.112)$
	2 param-s	0.657	0.785	1.084	$B_3(5.0034;4.109;9.161;2.343;0.073)$
2.0	Scale	0.980	1.203	1.771	$B_3(4.945;3.375;13.043;3.830;0.077)$
	form	0.922	1.125	1.625	$B_3(3.0434;4.162;7.1516;3.85;0.112)$
	2 param-s	0.643	0.766	1.051	$B_3(4.924;4.209;8.664;2.2754;0.073)$
3.0	Scale	0.952	1.163	1.702	$B_3(5.037;3.413;12.901;3.687;0.077)$
	form	0.912	1.110	1.601	$B_3(3.056;3.909;6.7844;3.397;0.112)$
	2 param-s	0.639	0.761	1.043	$B_3(4.9475;4.207;8.669;2.251;0.073)$
4.0	Scale	0.937	1.141	1.662	$B_3(4.943;3.504;12.224;3.630;0.077)$
	form	0.906	1.103	1.590	$B_3(3.053;3.944;6.762;3.3993;0.112)$
	2 param-s	0.637	0.758	1.039	$B_3(4.9274;4.228;8.557;2.239;0.073)$
5.0	Scale	0.927	1.130	1.640	$B_3(4.881;3.576;11.789;3.605;0.077)$
	form	0.902	1.099	1.586	$B_3(3.050;3.964;6.751;3.4024;0.112)$
	2 param-s	0.636	0.757	1.037	$B_3(4.921;4.243;8.488;2.2314;0.073)$

Table 9. Upper percentage points and models of limiting statistic distributions of the Cramer-Mises-Smirnov’s test when MLE are used in the case of gamma-distribution

Value of the form parameter	Parameter estimated	Percentage points			Model
		0.1	0.05	0.01	
0.3	Scale	0.233	0.305	0.482	$B_3(3.272;1.9595;16.177;0.75;0.013)$
	form	0.166	0.209	0.316	$B_3(3.025;3.226;11.11;0.776;0.0125)$
	2 param-s	0.127	0.158	0.233	$B_3(2.361;4.084;7.061;0.619;0.0145)$
0.5	Scale	0.205	0.264	0.413	$B_3(3.230;2.1984;14.315;0.70;0.013)$
	form	0.159	0.199	0.298	$B_3(3.014;3.35;10.095;0.721;0.0125)$
	2 param-s	0.119	0.146	0.212	$B_3(2.7216;3.984;7.499;0.537;0.013)$
1.0	Scale	0.175	0.220	0.336	$B_3(3.1201;2.546;11.120;0.60;0.013)$
	form	0.149	0.186	0.273	$B_3(2.993;3.472;8.828;0.635;0.0125)$
	2 param-s	0.111	0.136	0.194	$B_3(3.00;3.896;7.3247;0.4508;0.012)$
2.0	Scale	0.155	0.193	0.288	$B_3(2.9463;3.1124;9.116;0.60;0.013)$
	form	0.142	0.176	0.256	$B_3(2.991;3.533;8.20;0.5786;0.0125)$
	2 param-s	0.107	0.131	0.185	$B_3(3.053;3.940;7.117;0.425;0.0118)$
3.0	Scale	0.148	0.184	0.272	$B_3(2.884;3.3796;8.4342;0.60;0.013)$
	form	0.139	0.172	0.251	$B_3(2.974;3.553;7.884;0.555;0.0125)$
	2 param-s	0.106	0.129	0.182	$B_3(3.070;3.962;7.034;0.416;0.0117)$
4.0	Scale	0.145	0.179	0.264	$B_3(2.852;3.5285;8.1044;0.60;0.013)$
	form	0.138	0.171	0.248	$B_3(2.968;3.543;7.763;0.542;0.0125)$
	2 param-s	0.105	0.128	0.180	$B_3(3.097;3.954;7.064;0.412;0.0116)$
5.0	Scale	0.142	0.176	0.259	$B_3(2.8249;3.628;7.8756;0.60;0.013)$
	form	0.137	0.169	0.246	$B_3(2.964;3.547;7.656;0.533;0.0125)$
	2 param-s	0.105	0.128	0.179	$B_3(4.433;3.626;10.55;0.410;0.0084)$

V. CONCLUSION

In this paper we present more precise models of statistic distributions of the nonparametric goodness-of-fit tests in testing composite hypotheses subject to some laws considered in recommendations for standardization R 50.1.037-2002 [15]. The models of statistic distributions of the nonparametric goodness-of-fit tests for testing composite hypotheses of the Exponential family [14] were made more precise earlier and haven’t been improved now.

In the case of the I, II, III type beta-distribution families’ statistic distributions depend on a specific value of two form parameter of these distributions. Statistic distributions models and tables of percentage points for various combinations of values of two form parameters (more than 1500 models) were constructed in the thesis of Lemeshko S.B. [16] and partly were published in the paper [17].

The results of comparative analysis of goodness-of-fit tests power (nonparametric and χ^2 type) subject to some sufficiently close pair of alternative are presented in [18].

ACKNOWLEDGMENT

This research was supported by the Russian Foundation for Basic Research, project no. 06-01-00059a.

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