

PROBABILISTIC MODEL OF ALLOCATION LAWS OF EXPERIMENTAL DATA IN INFORMATION SYSTEMS

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ABSTRACT

On the basis of the beta distributions of the 1st and 2nd kind were received probabilistic models of distribution laws, which allow to approximate wider class of distribution laws of experimental data, than the existing Pearson's system of distributions. The method of identification parameters was developed of the generalized beta distribution using power, exponential and logarithmic moments.

Keywords: *Information System, Experimental Data, Probability, Distribution Laws, Approximation.*

1. INTRODUCTION

There is a view of the law of distribution assumed to be known in classical mathematical statistics and observing the results of its parameters assessed values. But usually pre-form of the distribution law is unknown and theoretical assumptions do not allow it to establish unequivocally. Also, processing of the experimental data does not allow to calculate accurately true distribution law. In this case, you should talk only about the approximation (approximate description) of real law to some others which is consistent with experimental data and in some ways similar to the unknown true law.

Nowadays, for the approximation of the experimental data distribution laws often used Pearson distributions [1-26]. However, the determination of the parameters of the desired distribution from the family of Pearson distributions connected with the decision of the various systems of equations using the method of moments. Besides, the method of moments does not allow to find the parameter estimates those distributions, including those owned by the Pearson family which do not have higher order moments (3rd and 4th). That is why the development of continuous distributions systems wider than the family of Pearson curves, as well as new methods for estimating the parameters has great importance both in theoretical and applied research.

Except of the method of Pearson for this purpose can be used the method based on obtaining a new distribution as a random function argument with the known distribution [2,6,8].

2. STATEMENT OF THE PROBLEM

The main objectives of the work:

- 1) To receive generalized beta distribution based on the beta distribution of the 1st and 2nd kind using method of functional transformation.
- 2) To consider the possibility of approximating the distribution law of experimental data, taking positive and negative values or only positive values using generalized beta distribution of the 1st and 2nd kind.

3. SOLUTION OF THE PROBLEM

3.1 Unilateral Generalized Beta Distributions Of 1st And 2nd Kind

Probability density functions (PDF) for the classical beta distributions of 1st and 2nd kind are as follows [2,27]:

$$p(y) = \frac{y^{\alpha-1}}{B(\alpha, \nu)} (1-y)^{\nu-1}, \quad 0 < y < 1; \quad (1)$$

$$p(y) = \frac{y^{\alpha-1}}{B(\alpha, \nu)(1+y)^{\alpha+\nu}}, \quad 0 < y < \infty, \quad (2)$$

where $\alpha > 0$, $\nu > 0$ - parameters of the form; $B(a, b)$ - beta function.

After a functional conversion $y = x^c / \chi^c$ or $y = \chi^c / x^c$ PDF (1) respectively, we have

$$p(x) = \frac{c x^{\alpha c - 1}}{B(\alpha, \nu) \chi^{\alpha c}} \left(1 - \frac{x^c}{\chi^c}\right)^{\nu - 1}, \quad 0 < x < \chi; \quad (3)$$

$$p(x) = \frac{c \chi^{\alpha c}}{B(\alpha, \nu) x^{\alpha c + 1}} \left(1 - \frac{\chi^c}{x^c}\right)^{\nu - 1}, \quad \chi < x < \infty, \quad (4)$$

where $\alpha > 0, \nu > 0, c > 0$ - parameters of the form; $\chi > 0$ - scale parameter.

Specific cases of distribution (3) there is the power law when $c = 1$ и $\nu = 1$; Beta distribution when $c = 1$. The limiting case of (3) is a lognormal distribution when $\alpha \rightarrow \infty, \nu \rightarrow \infty$ and $c \rightarrow 0$. A special case of PDF (4) is a Pareto distribution when $c = 1$ and $\nu = 1$ [27-29].

Using PDF (3) or PDF (4) and the ratio [30]

$$m_s = \int_0^\infty x^s p(x) dx, \quad (5)$$

We can get the initial moments of s -th order for distributions (3) and (4)

$$m_s = \frac{\chi^s \Gamma(\alpha + s/c) \Gamma(\alpha + \nu)}{\Gamma(\alpha) \Gamma(\alpha + \nu + s/c)};$$

$$m_s = \frac{\chi^s \Gamma(\alpha - s/c) \Gamma(\alpha + \nu)}{\Gamma(\alpha) \Gamma(\alpha + \nu - s/c)}, \quad (6)$$

where $\Gamma(z)$ - is the gamma function.

From (6) it follows that for PDF (4), there are only the initial direct points, order s of those satisfies the condition $s < \alpha c$. On the image 1 are presented the regions of existence of PDF(3) and PDF(4) in the plane of variables K_1 and K_2 , determined by the expressions (when $c = 1$) [31]

$$K_1 = \frac{m_{1c}^2}{m_{2c}}; \quad K_2 = \frac{m_{1c} m_{3c} - m_{2c}^2}{m_{2c} (m_{2c} - m_{1c}^2)}. \quad (7)$$

On the image 1 the region of distribution's existence (3) and its special cases left to the curve 3, characterizing a region of existence of the standard logarithmic distribution. Curve 1 characterizes the region of existence of the power law, and the point G - Gaussian distribution. The region of existence of the beta distribution is located to the left of the curve 2. The region of existence of distribution (4) and its special cases is located to the right of the curve 3. The curve 5 characterizes the region of the existence of Pareto distribution.

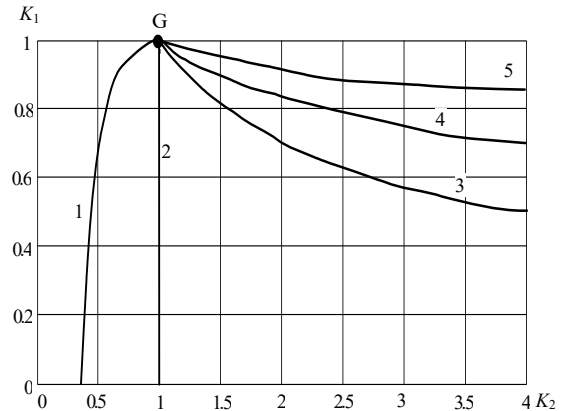


Figure 1. The diagram of unilateral laws of distribution

Performing the functional transformation $y = x^c / \lambda^c$ or $y = \lambda^c / x^c$ of PDF (2), respectively obtain

$$p(x) = \frac{c x^{\alpha c - 1}}{B(\alpha, \nu) \lambda^{\alpha c} \left(1 + \frac{x^c}{\lambda^c}\right)^{\alpha + \nu}}, \quad 0 < x < \infty, \quad (8)$$

$$p(x) = \frac{c \lambda^{\alpha c}}{B(\alpha, \nu) x^{\alpha c + 1} \left(1 + \frac{\lambda^c}{x^c}\right)^{\alpha + \nu}}, \quad 0 < x < \infty, \quad (9)$$

where $\nu > 0, 0 < \alpha \leq \nu, c > 0$ - parameters of the form; $\lambda > 0$ - scale parameter.

Special cases of PRV (8) are: beta distribution of II sort with $c = 1$; F- distribution when $\alpha = 0,5 n_1, \nu = 0,5 n_2, c = 1$ and $\lambda = n_2 / n_1$ [27].

After substituting the PDF (8) or PDF (9) in (5) and integration we obtain the early moments of s -th order for these distributions [32]

$$m_s = \frac{\lambda^s \Gamma(\alpha + s/c) \Gamma(\nu - s/c)}{\Gamma(\alpha) \Gamma(\nu)}. \quad (10)$$

$$m_s = \frac{\lambda^s \Gamma(\alpha - s/c) \Gamma(\nu + s/c)}{\Gamma(\alpha) \Gamma(\nu)}. \quad (11)$$

From (10) it follows that for PDF (8), there are only the initial moments, order of which s satisfies the condition $s < \nu c$. From (11) it follows that there are only the starting points for the PDF (9) order of which s satisfies the conditions $s < \alpha c$.

Let's consider the limiting case of PRV (3) and (8), when the parameter $\nu \rightarrow \infty$. In this case distributions (3) and (8) would be transformed into the generalized gamma distribution [29]

$$p(x) = \frac{c x^{\alpha c - 1}}{\Gamma(\alpha) \beta^{\alpha c}} \exp\left(-\frac{x^c}{\beta^c}\right), 0 < x < \infty, \quad (12)$$

where $\alpha > 0, c > 0$ – parameters of the form; $\beta > 0$ – scale parameter.

Special cases of (12) are: Rayleigh PDF when $\alpha = 1, \beta = \sqrt{2} \sigma$ and $c = 2$; exponential distribution with $\alpha = 1$ and $c = 1$; gamma distribution when $\alpha = \nu + 1, c = 1$; chi-square distribution when $\alpha = 0.5n, c = 1$ и $\beta = 2$; Nakagami distribution when $\alpha = m, \beta = \sqrt{\Omega/m}$ и $c = 2$; Weibull distribution when $\alpha = 1$ and $\beta^{-c} = \lambda$. Extreme cases (9) are the power law when $\alpha \rightarrow 0$ and $c \rightarrow \infty$; lognormal distribution when $\alpha \rightarrow \infty$ and $c \rightarrow 0$ [5-7].

Substituting the (12) into (5) and integrating [28] we obtain the initial moments of the s-th order

$$m_s = \beta^s \frac{\Gamma(\alpha + s/c)}{\Gamma(\alpha)}. \quad (13)$$

It should be noted, that the property, inherent in the distribution of (12) and presented in the form of equity [27]:

$$\frac{m_{(n+1)c} m_{1c} - m_{nc} m_{2c}}{(n-1)m_{nc}(m_{2c} - m_{1c}^2)} = 1, \quad (14)$$

when $n \geq 2$. This property is proved by substituting expression (13) into (14) for the corresponding initial moments.

On the image 1 the existence region of PRV (12) is located between the curves 1 and 3. Direct line 2 corresponds to the region of existence of the gamma distribution. The region of existence PDF(8) is located between curve 1 and curve 3. It is overlapped considerably with the region of existence of PDF (3) and includes a full region of existence of distribution (12).

Let's consider the limiting case of PRVDF (4) and PDF (9) when the parameter $\nu \rightarrow \infty$. In this case distributions (4) and (9) are converted into distribution

$$p(x) = \frac{c \beta^{\alpha c}}{\Gamma(\alpha) x^{\alpha c + 1}} \exp\left(-\frac{\beta^c}{x^c}\right), 0 < x < \infty. \quad (15)$$

where $\alpha > 0, c > 0$ – parameters of the form; $\beta > 0$ – scale parameter.

Special cases of (15) when $c = 1$ V is a type of distribution of the Pearson classification, and when $c \rightarrow \infty$ - is a Pareto distribution [27, 28]. Substituting the (15) in (5) and integrating, we obtain the initial moments of s-th order

$$m_s = \beta^s \Gamma(\alpha - s/c) / \Gamma(\alpha). \quad (16)$$

From (16) it follows that for the PRV (15), there are only the initial straight times, the order of which satisfies the condition $s < \alpha c$.

On the image 1 the existence region of PDF (15) is located between curves 3 and 5. The curve 4 describes a region of existence V-th type distribution of the Pearson classification. The region of existence of distribution (9) is located between the third curve and the fifth curve. It is overlapped considerably with the region of the existence of distribution (4) and includes a full region of existence of PDF (15).

For the obtained distributions (3), (8) and (12) are characterized by two properties:

1) the property of moments defined by the equation:

$$\frac{m_{3c}}{m_{4c}} \cdot \frac{3m_{2c}m_{3c} - m_{1c}(4m_{1c}m_{3c} - m_{2c}^2)}{4m_{2c}^2 - m_{1c}(3m_{1c}m_{2c} + m_{3c})} = 1. \quad (17)$$

2) The condition of distribution (3) is $K_2 < 1$, for distributing (12) - $K_2 = 1$ and for distributing (8) - $K_2 > 1$.

These properties are also valid for the distribution (4), (9) and (15), if to substitute in relations (7) and (17) instead of direct power moments inverse points. In this case for distribution (4) is still the condition $K_2 < 1$, for distributing (15) - $K_2 = 1$ and for distributing (9) - $K_2 > 1$. These properties can be used to identify the generalized beta distribution.

Let us to consider now the limiting cases for generalized beta distribution of the 1st and 2nd kind, when the parameter $c \rightarrow 0$. In this case distribution (12) and (15) are converted into logarithmic normal distribution. The expression for the PDF has the form

$$p(x) = \frac{1}{\sqrt{2\pi} \sigma x} \exp\left(-\frac{(\ln(x) - \mu)^2}{2\sigma^2}\right), \quad (18)$$

where $\mu > 0, \sigma > 0$ - distribution options.

For the distribution of (18) there are all direct and inverse power moments. Therefore PDF parameters are defined by the first initial and the second central moments of the logarithmic

$$\hat{\mu} = \hat{l}_1, \hat{\sigma} = \sqrt{\hat{L}_2}. \quad (19)$$

The region of existence PRV (18) is shown at the image 1 by the curve 3.

Distribution (3) and (8) converted into distribution, when the parameter $c \rightarrow 0$

$$p(x) = \frac{\beta^\nu x^{\beta-1}}{\Gamma(\nu) \chi^\beta} \left(\ln\left(\frac{\chi}{x}\right)\right)^{\nu-1}, 0 < x < \chi; \quad (20)$$

where $\nu > 0, \beta > 0$ – parameters of the form; $\chi > 0$ – scale parameter.

Substituting the PDV (20) in equation (5) and integrating [32], we get the early moments of the s -th order

$$m_s = (\beta/(\beta + s))^\nu \chi^s. \quad (21)$$

On the image 1 the existence region of distribution (20) is located to the left of the curve 3.

When $c \rightarrow 0$ distributions (4) and (9) are converted into distribution

$$p(x) = \frac{\beta^\nu \chi^\beta}{\Gamma(\nu) x^{\beta+1}} \left(\ln \left(\frac{x}{\chi} \right) \right)^{\nu-1}, \quad \chi < x < \infty, \quad (22)$$

where $\nu > 0, \beta > 0$ – parameters of the form; $\chi > 0$ – scale parameter.

Substituting the GHD (22) into (5) and integrating [32], We obtain the initial moments of s -th order

$$m_s = (\beta/(\beta - s))^\nu \chi^s. \quad (23)$$

From (23) it follows that for distributing (22) there are all inverse initial moments and only those points straight, order of which satisfies $s < \beta$. On the image 1 the region of distribution's existence (22) is located to the right of the curve 3. The distribution parameters (22) are defined by (23) with the help of reverse moments.

Identification of the lognormal distribution (18) is possible only with the use of logarithmic points [27]. Except of PDF (18), this group includes the distribution

$$p(x) = \frac{(-0,5 \ln(h))^{1-2\nu}}{B(0,5, \nu) x} \left(\ln \left(\frac{x}{h\chi} \right) \ln \left(\frac{\chi}{x} \right) \right)^{\nu-1}, \quad h\chi < x < \chi; \quad (24)$$

$$p(x) = \frac{\lambda^{2\nu}}{B(0,5, \nu) x \left[\lambda^2 + (\ln(x) - \mu)^2 \right]^{\nu+0,5}}, \quad 0 < x < \infty. \quad (25)$$

where $0 < h < 1$ – the form parameter.

They are also limiting distributions for generalized beta distribution. There is only a portion of the logarithmic moments existing for PDF (25). Thus, we get a wide class of models of unilateral laws of distributions based on beta-distributions of the 1st and 2nd kind. When we identifying generalized beta distribution taking into account the consideration of their properties, you can use the forward and reverse power moments (including fractional order), and logarithmic moments [31,33,34].

3.2 Identification Of Unilateral Generalized Beta Distributions Of The 1st And 2nd Kind

Approximation of the experimental distributions with the help of unilateral generalized beta

distribution can be carried out using the following algorithm:

1. Initially are determining logarithmic sampling points

$$\hat{l}_1 = \frac{1}{n} \sum_{i=1}^n \ln x_i,$$

$$\hat{L}_s = \frac{1}{n} \sum_{i=1}^n (\ln(x_i) - \hat{l}_1)^s, \quad s = 2, 3, \quad (26)$$

and then an estimate the asymmetry coefficient

$$\hat{L}_a = \hat{L}_3 / \hat{L}_2^{1,5}.$$

2. If for the coefficient \hat{L}_a the following condition is $-0,1 \leq \hat{L}_a \leq 0,1$, then further defined selective central point \hat{L}_4 and is a joint evaluation of coefficient skewness and kurtosis

$$\hat{L}_{ae} = \frac{6\hat{L}_2^2}{\hat{L}_4 + 3\hat{L}_2^2}.$$

When the condition $\hat{L}_{ae} > 1,04$ to approximate of the experimental distribution is used the distribution (24) with parameters

$$\hat{\nu} = \frac{1,5 - \hat{L}_{ae}}{\hat{L}_{ae} - 1};$$

$$\hat{\chi} = \exp \left(\sqrt{\frac{2 - \hat{L}_{ae}}{\hat{L}_{ae} - 1}} \hat{L}_2 + \hat{l}_1 \right);$$

$$\hat{h} = \exp \left(-2 \sqrt{\frac{2 - \hat{L}_{ae}}{\hat{L}_{ae} - 1}} \hat{L}_2 \right).$$

If for the coefficient \hat{L}_a the following condition is $0,96 \leq \hat{L}_a \leq 1,04$, the lognormal distribution (22) is used to approximate the experimental distribution, whose parameters according to (23) defined by the initial first and second central logarithmic moments ($\hat{\mu} = \hat{l}_1, \hat{\sigma} = \sqrt{\hat{L}_2}$).

When the condition $\hat{L}_{ae} < 0,96$ to approximate the experimental distribution is used the distribution (25) with parameters

$$\hat{\nu} = \frac{2 - 1,5\hat{L}_{ae}}{1 - \hat{L}_{ae}}; \quad \hat{\lambda} = \sqrt{\frac{2 - \hat{L}_{ae}}{1 - \hat{L}_{ae}}} \hat{L}_2; \quad \hat{\mu} = \hat{l}_1.$$

3. If the ratio $\hat{L}_a < -0,1$, then for the approximation of the experimental distribution is used one of the distributions (3), (8) or (12). Type of the distribution and its parameters can be

determined as follows: first estimate is the parameter c of the solution of equation

$$\frac{\hat{m}_{3c}}{\hat{m}_{4c}} \cdot \frac{3\hat{m}_{2c}\hat{m}_{3c} - \hat{m}_{1c}(4\hat{m}_{1c}\hat{m}_{3c} - \hat{m}_{2c}^2)}{4\hat{m}_{2c}^2 - \hat{m}_{1c}(3\hat{m}_{1c}\hat{m}_{2c} + \hat{m}_{3c})} = 1, \quad (27)$$

where

$$\begin{aligned} \hat{m}_{1c} &= \frac{1}{n} \sum_{i=1}^n x_i^{\hat{c}}, \quad \hat{m}_{2c} = \frac{1}{n} \sum_{i=1}^n x_i^{2\hat{c}}, \\ \hat{m}_{3c} &= \frac{1}{n} \sum_{i=1}^n x_i^{3\hat{c}}, \quad \hat{m}_{4c} = \frac{1}{n} \sum_{i=1}^n x_i^{4\hat{c}}. \end{aligned} \quad (28)$$

Then, the coefficient's estimates are determined as K_1 и K_2 with help of the relations (7). If the condition $\hat{K}_2 < 1$, then for the approximation of the experimental distribution, use distribution (3) with parameters

$$\begin{aligned} \hat{\alpha} &= \frac{2\hat{K}_1\hat{K}_2}{1 + \hat{K}_2 - 2\hat{K}_1\hat{K}_2}; \\ \hat{\nu} &= \frac{2\hat{K}_2}{1 - \hat{K}_2} - \hat{\alpha}; \quad \hat{\chi} = \left(\hat{m}_{1c} \left(1 + \frac{\hat{\nu}}{\hat{\alpha}} \right) \right)^{1/\hat{c}}. \end{aligned} \quad (29)$$

If $\hat{K}_2 = 1$, then to approximate the experimental distribution is used the distribution (12) with parameters

$$\hat{\alpha} = \frac{\hat{K}_1}{1 - \hat{K}_1}; \quad \hat{\beta} = \left(\frac{\hat{m}_{1c}}{\hat{\alpha}} \right)^{1/\hat{c}}. \quad (30)$$

When the condition is $\hat{K}_2 > 1$ to approximate the experimental distribution is used the distribution (8) with the following parameters

$$\begin{aligned} \hat{\alpha} &= \frac{2\hat{K}_1\hat{K}_2}{1 + \hat{K}_2 - 2\hat{K}_1\hat{K}_2}; \\ \hat{\nu} &= \frac{2\hat{K}_2}{\hat{K}_2 - 1} + 1; \quad \hat{\lambda} = \left(\hat{m}_{1c} \frac{\hat{\nu} - 1}{\hat{\alpha}} \right)^{1/\hat{c}}. \end{aligned} \quad (31)$$

4. If the ratio $\hat{L}_a > 0,1$, then for the approximation of the experimental distribution is used one of the distributions (4), (9) or (15). Type of distribution and its parameters can be determined as follows: first evaluate determined parameter c by solving the equation (27), and then determine the coefficient's estimates K_1 и K_2 with help of the relations (7). In this case, relations (7) and (17) are now used selective inverse points

$$\begin{aligned} \hat{m}_{1c} &= \frac{1}{n} \sum_{i=1}^n x_i^{-\hat{c}}, \quad \hat{m}_{2c} = \frac{1}{n} \sum_{i=1}^n x_i^{-2\hat{c}}, \\ \hat{m}_{3c} &= \frac{1}{n} \sum_{i=1}^n x_i^{-3\hat{c}}, \quad \hat{m}_{4c} = \frac{1}{n} \sum_{i=1}^n x_i^{-4\hat{c}}. \end{aligned} \quad (32)$$

If the condition $\hat{K}_2 < 1$, you should use the distribution (4) for the approximation of the experimental distribution with parameters $\hat{\alpha}$, ν and $\hat{\chi}$, which are defined by (29) with considering the (32).

If $\hat{K}_2 = 1$, then to approximate the experimental distribution the distribution (15) with parameters $\hat{\alpha}$ and $\hat{\beta}$ are used, defining relations (30) with considering the (32).

When the condition $\hat{K}_2 > 1$ is true for approximating the experimental distribution is used the distribution (9) with parameters (31)

5. When the parameter's estimate $\hat{c} \rightarrow 0$ (on practice $\hat{c} \leq 0,1$), then the approximation of the experimental distributions is made with distributions (20) and (22). If the ratio $\hat{L}_a < -0,1$, It is used PDF (20) to approximate the distribution. Its parameters can be determined as follows: first find the estimate of the parameter β is determined by solving the equation

$$\frac{\ln(\hat{m}_1) - \hat{l}_1}{\ln(\hat{m}_2) - 2\ln(\hat{m}_1)} = \frac{\ln(\hat{\beta}) - \ln(\hat{\beta} + 1) + 1/\hat{\beta}}{2\ln(\hat{\beta} + 1) - \ln(\hat{\beta}) - \ln(\hat{\beta} + 2)},$$

where

$$\hat{m}_1 = \frac{1}{n} \sum_{i=1}^n x_i, \quad \hat{m}_2 = \frac{1}{n} \sum_{i=1}^n x_i^2.$$

Then, evaluation parameters are ν and χ determined with the aid of relations

$$\hat{\nu} = \frac{\ln(\hat{m}_1) - \hat{l}_1}{\ln(\hat{\beta}) - \ln(\hat{\beta} + 1) + 1/\hat{\beta}}, \quad \hat{\chi} = \exp\left(\hat{l}_1 + \frac{\hat{\nu}}{\hat{\beta}}\right).$$

If the ratios $\hat{L}_a > 0,1$ and $\hat{c} \leq 0,1$, then is used PDF (22) for approximating the distribution. It's parameters are defined in a similar distribution of parameters (20). The estimate of parameter β is determined by solving the equation

$$\frac{\ln(\hat{m}_1) + \hat{l}_1}{\ln(\hat{m}_2) - 2\ln(\hat{m}_1)} = \frac{\ln(\hat{\beta}) - \ln(\hat{\beta} + 1) + 1/\hat{\beta}}{2\ln(\hat{\beta} + 1) - \ln(\hat{\beta}) - \ln(\hat{\beta} + 2)},$$

where

$$\hat{m}_1 = \frac{1}{n} \sum_{i=1}^n \frac{1}{x_i}, \quad \hat{m}_2 = \frac{1}{n} \sum_{i=1}^n \frac{1}{x_i^2}.$$

Parameter's estimates ν and χ corresponding expression

$$\hat{\nu} = \frac{\ln(\hat{m}_1) + \hat{l}_1}{\ln(\hat{\beta}) - \ln(\hat{\beta} + 1) + 1/\hat{\beta}}, \quad \hat{\chi} = \exp\left(\hat{l}_1 - \frac{\hat{\nu}}{\hat{\beta}}\right).$$

If the resulting parameter estimate $\hat{c} \geq 3$, it can be assumed that $\hat{c} = 3$. The error of approximation of the experimental distribution increases slightly.

Our procedure approximation of the experimental distributions should be used when sample size of $n \geq 1000$.

Similarly it is possible to carry out an approximation of the theoretical distributions, but instead of the sample moments in this case use the appropriate power and logarithmic points of approximating the theoretical distribution.

3.3 Bilateral Generalized Beta Distribution And Their Identification

Bilateral generalized beta distribution can be obtained by functional transformation $z = \ln(x)$ unilateral generalized beta distribution of the 1st and 2nd kind. As a result, the functional transformation of the distributions (3), (4), (8), (9), (12), (15), (18), (20), (22), (24) and (25) will take correspondingly the following form:

$$p(z) = \frac{c \exp(\alpha c(z-\mu))}{B(\alpha, v)} (1 - \exp(c(z-\mu)))^{v-1}, -\infty < z < \mu; (33)$$

$$p(z) = \frac{c \exp(-\alpha c(z-\mu))}{B(\alpha, v)} (1 - \exp(-c(z-\mu)))^{v-1} \quad (34)$$

$$p(z) = \frac{c \exp(\alpha c(z-\mu))}{B(\alpha, v) [1 + \exp(c(z-\mu))]^{\alpha+v}}, \mu < z < \infty; -\infty < z < \infty; (35)$$

$$p(z) = \frac{c \exp(-\alpha c(z-\mu))}{B(\alpha, v) [1 + \exp(-c(z-\mu))]^{\alpha+v}}, \mu < z < \infty; (36)$$

$$p(z) = \frac{c \exp(\alpha c(z-\mu))}{\Gamma(\alpha)} \exp[-\exp(c(z-\mu))], -\infty < z < \infty; (37)$$

$$p(z) = \frac{c \exp(-\alpha c(z-\mu))}{\Gamma(\alpha)} \exp[-\exp(-c(z-\mu))], -\infty < z < \infty; (38)$$

$$p(z) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left[-\frac{(z-\mu)^2}{2\sigma^2}\right], -\infty < z < \infty; (39)$$

$$p(z) = \frac{\lambda^\alpha}{\Gamma(\alpha)} (\mu-z)^{\alpha-1} \exp[-\lambda(\mu-z)], -\infty < z < \mu; (40)$$

$$p(z) = \frac{\lambda^\alpha}{\Gamma(\alpha)} (z-\mu)^{\alpha-1} \exp[-\lambda(z-\mu)], \mu < z < \infty; (41)$$

$$p(z) = \frac{(z+\chi-\mu)^{v-1} (\chi-z+\mu)^{v-1}}{(\chi)^{2v-1} B(0.5, v)}, -\chi+\mu < x < \chi+\mu; (42)$$

$$p(z) = \frac{\lambda^{2v}}{B(0.5, v) [\lambda^2 + (z-\mu)^2]^{v+0.5}}, -\infty < z < \infty; (43)$$

where μ - shift parameter. In (35) and (36) satisfies the condition $0 < \alpha \leq v$.

Bilateral generalized beta distribution of the 1st and 2nd kind (33)-(43) can be used for the approximation of the experimental distributions NE, taking negative and positive values. When defining their parameters instead of random direct and inverse power moments using direct and inverse exponential moments, and selective direct power points instead of logarithmic moments. This allows you to apply for identification of the parameters of bilateral generalized beta distribution algorithm, discussed above in Section 2. It is found that if the ratio $\hat{L}_a < 0$, then in the law of distribution are dominating direct exponential moments, and if $\hat{L}_a > 0$, the predominant circulating exponential moments. Similarly it is possible to carry out an approximation of the theoretical distributions of bilateral, but instead of sampling points used in this case the corresponding power and exponential moments approximating the theoretical distribution are used.

4. CONCLUSIONS

Thus, there were proposed generalized beta distribution to approximate the laws of unilateral and bilateral distribution of experimental data. This allows to receive wider class of distributions laws, than the existing system of Pearson distributions. Was developed a method for identifying the parameters of generalized beta distribution of the 1st and 2nd kind with the use of power, exponential and logarithmic moments. In this case it is possible in many cases to increase the accuracy of the parameter's estimates of the distributions. A topographic classification of unilateral distribution laws was developed.

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