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Opinion

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The New Size-Biased Kumaraswamy-G Distribution

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Abstract

This paper introduces a new size-biased distribution which is partially inspired by [1]. A member of this new class of distributions is shown to be practically significant in modeling real life data. Finally, we ask the reader to investigate further properties and applications of this new class of

Keywords: Size biased distribution; Regularized beta function; Normal distribution

The Size-Biased Kumaraswamy Distribution

The concept of weighted distributions is largely credited to Fisher [1,2] and Rao [3]. In the size biased schemes, the probability of sampling an individual is proportional to X^r provided that $E_{\theta}(X^r) < \infty$ for all θ . In such situations, the weighted probability density function is defined as

$$f_r(x;\theta) = \frac{x^r f(x;\theta)}{\mu'_r}$$

where $\mu'_r = \int x^r f(x;\theta) dx$. When the probability of observing a positive-valued random variable is proportional to the value of the variable the resulting distribution is termed size-biased with probability density function

$$g(x;\theta) = \frac{xf(x;\theta)}{\mu'_r}$$

The Kumaraswamy distribution [4] has probability density function

$$g(x;a,b) = abx^{a-b} (1-x^a)^{b-1}, 0 < x < 1; a,b > 0$$

and the rth order raw moment is given by [1]

$$\mu'_r = bB\left(1 + \frac{r}{a}, b\right)$$

 $\mu_r' = bB\bigg(1+\frac{r}{a},b\bigg)$ where $B\bigg(1+\frac{r}{a},b\bigg)$ is a beta function defined by the integral

$$B(\alpha,\beta) = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx$$

Notation 1.1. We write $X \sim SBKD(a,b)$, if X is a size biased Kumaraswamy random variable

Proposition 1.2. [1] The PDF of the SBKD(a, b) distribution is given by

$$f(x;a,b) = \frac{ax^{a-1}(1-x^a)^{b-1}}{B(1+\frac{1}{a},b)}$$

where 0 < x < 1, and $B\left(1 + \frac{r}{a}, b\right)$ is a beta function defined by the integral

$$B(\alpha, \beta) = \int_0^1 x^{\alpha - 1} (1 - x)^{\beta - 1} dx$$

Theorem 1.3. [1] The CDF of the SBKD (a, b) distribution is given by

$$F(x;a,b) = I_{x^a} \left(1 + \frac{1}{a}, b\right)$$

Where

$$I_{x^a}\left(1+\frac{1}{a},b\right) == \frac{B\left(x^a;1+\frac{1}{a},b\right)}{B\left(1+\frac{1}{a},b\right)}$$

is the regularized incomplete beta function and is defined as the ratio of an incomplete beta function,

$$B(z;\alpha,\beta) = \int_0^z x^{\alpha-1} (1-x)^{\beta-1} dx$$

and the complete beta function,

$$B(\alpha,\beta) = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx$$

with
$$a, b > 0$$
, and $x \in (0,1)$

The $\left(\frac{1}{e}\right)^{\alpha}$ PT-G Family of Distributions

Definition 2.1. [5] Let $\alpha \neq \frac{1}{e}, \alpha > \frac{1}{e}$ and $\xi > 0$ where α is the rate scale parameter and is a vector of parameters in the baseline distribution all of whose entries are positive.

A random variable Z is said to follow the $\left(\frac{1}{a}\right)^a$ power transform family of distributions if

the Cumulative Distribution Function (CDF) is given by

$$F(x;\alpha,\xi) = \frac{1 - e^{-\alpha G(x;\xi)}}{1 - e^{-\alpha}} \qquad x \in \mathbb{R},$$

where the baseline distribution has CDF $G(x;\xi)$.

Proposition 2.2. [5] The PDF of the $\left(\frac{1}{e}\right)^{\alpha}$ power transform family distribution is given by

$$f(x;\alpha,\xi) = \frac{1}{1 - e^{-\alpha}} \left(\alpha g(x;\xi) e^{-\alpha G(x;\xi)} \right)$$

where $\alpha \neq \frac{1}{e}$, $\alpha > \frac{1}{e}$ and $\xi > 0$ and $G(x;\xi)$ is the cumulative distribution with

probability density function $g(x;\xi)$

The New Distribution

By modifying the parameter space for α in the $\left(\frac{1}{e}\right)^{\alpha}$ PT-Standard Uniform family and using it in Theorem 1.3, we have the following

Proposition 3.1. The CDF of the new size biased Kumaraswamy Distribution is given by

$$F(x;a,b,\alpha) = I_{\left(\frac{1-e^{-ax}}{1-e^{-ax}}\right)^a} \left(1 + \frac{1}{a},b\right)$$

$$I_{\left(\frac{1-e^{-ax}}{1-e^{-a}}\right)^{a}}\left(1+\frac{1}{a},b\right) = \frac{B\left(\left(\frac{1-e^{-ax}}{1-e^{-a}}\right)^{a};1+\frac{1}{a},b\right)}{B\left(1+\frac{1}{a},b\right)}$$

is the regularized incomplete beta function and is defined as the ratio of an incomplete beta function,

$$B(\alpha,\beta) = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx$$

and the complete beta function,

$$B(\alpha,\beta) = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx$$

with $a, b > 0, \alpha \in \mathbb{R}, \alpha \neq 0$, and $x \in (0,1)$

Using the above, we introduce the following

Definition 3.2. We say Z is a new size-biased Kumaraswamy-G random variable, and write $Z \sim SBKGD(a, b, \alpha, \xi)$, if the CDF is given by

$$F^*(x;a,b,\alpha,\xi) := F(G(x;\xi),a,b,\alpha)$$

where F is given by the previous Proposition, a, b > 0, $\alpha \in R$, $\alpha \in R$ $0, x \in R$, and ξ is a vector of parameters in the baseline distribution with CDF $G(x;\xi)$

Remark 3.3. The PDF of the new size-biased Kumaraswamy-G family of distributions can be obtained by differentiating the CDF

Practical Illustration

We assume the baseline distribution is Normal with the following CDF

$$G(x;c,d) = \frac{1}{2} erfc \left(\frac{c-x}{\sqrt{2}d}\right)$$

where $x, c \in \mathbb{R}$, d>0, and

$$\operatorname{erfc}(z) = 1 - \operatorname{erfc}(z) = 1 - \frac{2}{\sqrt{\pi}} \int_0^z e^{-t^2} dt.$$

Thus from Definition 3.2 and Proposition 3.1, we have the following

Proposition 4.1. The CDF of the new size-biased Kumaraswamy-Normal distribution is given by

$$F^*\left(x;a,b,\alpha,c,d\right) = I_{\left(\frac{1-e^{-\frac{1}{2}\alpha erf\left(\frac{c-x}{\sqrt{2}d}\right)}}{1-e^{-\alpha}}\right)}^a \left(\frac{1}{a} + 1,b\right)$$

Where

$$I_{z}(a,b) = \frac{B(z;a,b)}{B(a,b)} = \frac{\int_{0}^{z} x^{\alpha-1} (1-x)^{\beta-1} dx}{\int_{0}^{1} x^{\alpha-1} (1-x)^{\beta-1} dx}$$

with
$$x,c \in \mathbb{R},d>0$$
, and $\operatorname{erfc}(z) = 1 - \operatorname{erfc}(z) = 1 - \frac{2}{\sqrt{\pi}} \int_0^z e^{-t^2} dt$

Remark 4.2. We write Q ~ N SBKN D(a, b, α , c, d), if Q is a new size-biased Kumaraswamy-Normal random variable. When we fix a = 1, we denote the reduced distribution as N BIN D(b, α , c, d), and say Q is a new Beta-I-Normal random variable (Figure 1).

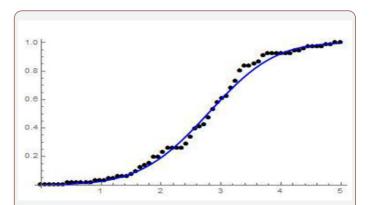


Figure 1: The CDF of COF W (0.73884, 1.47816, 0.192653, 2062.69) fitted to the em-pirical distribution of the data on patients with breast cancer, Section 5.2.

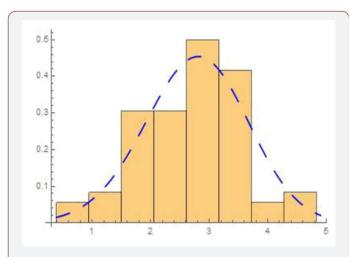


Figure 2: The PDF of N BIN D(3.63914, -0.194438, 3.33402, 1.53905) fitted to the histogram of Table 2[6].

Remark 4.3. The PDF of the new size-biased Kumaraswamy-Normal distribution can be obtained by differentiating the CDF (Figure 2).

Conclusion

Size-biased distributions arise naturally in a range of sampling and modeling problems in forestry [7], as such we suggest to the reader to obtain some properties and applications of this new class of distributions as it pertains to forestry.

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None.

Conflict of Interest

No conflict of interest.

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