# A Note on Alpha-Skew-Generalized Logistic Distribution

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**Abstract:** In this short article alpha-skew-generalized logistic distribution of type III has been introduced as an extension of the alpha-skew-logistic distribution [Hazarika and Chakraborty (2014). Alpha Skew Logistic Distribution, IOSR Journal of Mathematics 10 (4) Ver. VI: 36-46] by considering the generalized logistic distribution of type III as the base distribution. Cumulative distribution function, moment generating function and a few moments have been derived.

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#### I. Introduction

Skew distribution is natural extension of the underlying symmetric distribution derived by adding one or more additional asymmetry parameter(s). Azzalini [1] first introduced the path breaking skew-normal distribution whose probability density function (pdf) is given by

$$f_Z(z;\lambda) = 2\phi(z)\Phi(\lambda z); \qquad z \in R, \ \lambda \in R$$
 (1)

where  $\phi(.)$  and  $\Phi(.)$  are the pdf and cumulative distribution function (cdf) of standard normal distribution respectively. Here,  $\lambda$  is the asymmetry parameter. The general formula for the construction of skew-symmetric distributions was given by Huang and Chen [2] by introducing the concept of skew function G(.). According to Huang and Chen [2] a random variable Z is said to be skew symmetric if its pdf is given by

$$f_Z(z) = 2h(z)G(z); \ z \in R \tag{2}$$

where, G(.) (skew function) is the Lebesgue measurable function such that,  $0 \le G(z) \le 1$  and G(z) + G(-z) = 1;  $z \in R$ , almost everywhere and h(.) is any symmetric (about 0) pdf. One can construct many skew distributions showing both unimodal and multimodal behavior by choosing an appropriate skew function G(z) in the equation (2) (for more about skew distributions see [3]).

Alpha–skew–normal (ASN) distribution was introduced by Elal-Olivero [4] as a new class of skew normal distribution that includes unimodal as well as bimodal normal distributions. A random variable Z is said to be alpha–skew–normal distribution if its pdf is given by

$$f_{ASN}(z;\alpha) = \frac{\{(1-\alpha z)^2 + 1\}}{2+\alpha^2} \phi(z); -\infty < z < \infty, \ \alpha \in R$$

$$(3)$$

where,  $\phi(z)$  is the pdf of standard normal distribution. The generalized version of the ASN distribution has been introduced by Handam [5]. Using exactly the similar approach Harandi and Alamatsaz [6] investigated a class of alpha–skew–Laplace (ASL) distribution. Recently Hazarika and Chakraborty [7] introduced and studied many properties along with estimation of parameters and data fitting application of the alpha–skew–Logistic distribution (ASLG) having pdf

$$f_Z(z;\alpha) = \frac{\{(1-\alpha z)^2 + 1\}\exp(-z)}{C\{1+\exp(-z)\}^2}, z \in R, \ \alpha \in R$$
 (5)

where,  $C = 2 + \frac{\alpha^2 \pi^2}{3}$ . Symbolically if Z is an alpha-skew-logistic random variable with parameter  $\alpha$  it is denoted in this article by  $Z \sim \text{ASLG}(\alpha)$ .

In this note, alpha-skew-generalized logistic (ASGL) distribution has been introduced by considering the generalized logistic distribution of type III as the base distribution following the methodology of Elal-Olivero [4] and some of its properties have been studied.

### II. Alpha-Skew-Generalized Logistic Distribution

**Definition 2.1** A random variable Z is said to follow generalized logistic distribution type III (Balkrishnan, [8]; Jhonson et al., [9]) if its pdf is given by

$$f_{GL}(z) = \frac{1}{B(\beta, \beta)} \frac{\exp(-\beta z)}{\{1 + \exp(-z)\}^{2\beta}}; -\infty < z < \infty, \ \beta > 0$$
 (6)

where  $B(\beta, \beta) = \frac{\{\Gamma(\beta)\}^2}{\Gamma(2\beta)}$  and it is denoted by  $Z \sim GL(\beta)$ .

**Definition 2.2** A random variable Z is said to be alpha-skew-generalized logistic ASGLG( $\alpha, \beta$ ) if its pdf is given by

$$f_{\text{ASLG}}(z;\alpha) = \frac{(1-\alpha z)^2 + 1}{D} \frac{\exp(-\beta z)}{B(\beta,\beta)(1+\exp(-z))^{2\beta}}; \ -\infty < z < \infty, \ \beta > 0, \alpha \in R$$
 (7)

where,  $D = 2 [1 + \alpha^2 \psi^{(1)}(\beta)]$  and

$$\psi^{(n)}(a) = \frac{d^n}{d \, a^n} \Gamma(a) = \sum_{k=0}^{\infty} \frac{(-1)^{n+1} n!}{(a+k)^{n+1}}, \text{ for any positive integer } n \text{ is the poly gamma function (Gradshteyn)}$$

and Ryzhik, [10])

#### Particular cases:

- I.  $ASGLG(\alpha,1) = ASLG(\alpha)$  (Hazarika and Chakraborty [7])
- II. ASGLG $(0, \beta) = GL(\beta)$ .
- III. ASGLG(0,1) = ASLG(0) = L(0,1), where, L(0,1) is the standard logistic distribution and its pdf is given

by 
$$f(z) = \frac{\exp(-z)}{\{1 + \exp(-x)\}^2}$$
,  $-\infty < z < \infty$ .

IV. If  $Z \sim ASGLG(\alpha, \beta)$  then  $-Z \sim ASGLG(-\alpha, \beta)$ .

**Remark 2.1:** The Cumulative distribution function (cdf) of ASGLG( $\alpha, \beta$ ) is given by

$$F(z) = \frac{1}{D B(\beta, \beta)} \Big[ \beta^{-3} \exp(-\beta z) \{1 + \exp(-z)\}^{-2\beta} \{1 + \exp(z)\}^{2\beta} \Big\{ \beta^2 (2 + \alpha z(-2 + \alpha z)) + 2\beta (\beta, 2\beta, 1 + \beta, -\exp(z)) + 2\alpha \beta (-1 + \alpha z) \} \Big\} \Big]$$

$$= \frac{1}{D B(\beta, \beta)} \Big[ \beta^{-3} \exp(-\beta z) \{1 + \exp(-z)\}^{-2\beta} \{1 + \exp(z)\}^{-2\beta} \Big\} \Big]$$

where,  $_pF_q(a;b;x)$  is the generalized Hypergeometric function (Gradshteyn and Ryzhik, [10]) and given by

$$_{p}F_{q}(a;b;x) = \sum_{k=0}^{\infty} \frac{(a_{1})_{k}(a_{2})_{k}...(a_{p})_{k}}{(b_{1})_{k}(b_{2})_{k}...(b_{p})_{k}} \frac{x^{k}}{k!} \operatorname{and}(g)_{k} = g(g+1)\cdots(g+k-1)$$

**Proposition 2.1.** (Pogány and Nadarajah [11]) If  $Z \sim ASGLG(\alpha, \beta)$  then

$$E(Z^{2n-1}) = 0$$

$$E(Z^{2n}) = \frac{2^{n+1} (n!)^2}{B(\beta,\beta)} \Phi_{2u}^* (-1,2n+1,\beta) ; \forall n > \beta.$$

where,  $\Phi_{\mu}^{*}(z,s,a) = \sum_{j=0}^{\infty} \frac{(\mu)_{j} z^{j}}{(a+j)^{s} j!}$  is the Goyal–Laddha generalized Hurwitz–Lerch Zeta function

(see Goyal and Laddha [12])

**Result 2.1** If  $Z \sim ASGLG(\alpha, \beta)$  then

$$E(Z^{n}) = \frac{1}{DB(\beta,\beta)} \begin{cases} -2^{\frac{n+5}{2}} \alpha^{2} \left( ((n+1)/2)! \right)^{2} \Phi_{2u}^{*}(-1,n+2,\beta); & \text{if } n(>2\beta-1) \text{ is even} \\ 2^{\frac{n+4}{2}} \left[ \left( (n/2)! \right)^{2} \Phi_{2\beta}^{*}(-1,n+1,\beta) + \alpha^{2} \left( ((n+1)/2)! \right)^{2} \Phi_{2\beta}^{*}(-1,n+3,\beta) \right] \\ & \text{if } n(>\beta) \text{is odd} \end{cases}$$

Proof: 
$$E(Z^n) = \int_{-\infty}^{\infty} Z^n \frac{(1 - \alpha z)^2 + 1}{D} \frac{\exp(-\beta z)}{B(\beta, \beta)(1 + \exp(-z))^{2\beta}} dz$$

$$= \frac{1}{D} \left[ \int_{-\infty}^{\infty} 2z^n f_{GL}(z) dz + \int_{-\infty}^{\infty} \alpha^2 z^{n+2} f_{GL}(z) dz - 2 \int_{-\infty}^{\infty} \alpha^2 z^{n+1} f_{GL}(z) dz \right]$$

Now using Proposition 2.1 we get for the case of when n is odd

$$E(Z^{n}) = \frac{1}{D} \left[ -2 \int_{-\infty}^{\infty} \alpha^{2} z^{n+1} f_{GL}(z) dz \right]$$
$$= \frac{-2^{\frac{n+5}{2}} \alpha^{2} \left( \frac{(n+1)}{2}! \right)^{2}}{DB(\beta, \beta)} \Phi_{2u}^{*}(-1, n+2, \beta)$$

and for the case when n is even is obtained as

$$E(Z^{n}) = \frac{1}{D} \left[ \int_{-\infty}^{\infty} 2z^{n} f_{GL}(z) dz + \int_{-\infty}^{\infty} \alpha^{2} z^{n+2} f_{GL}(z) dz \right]$$

$$= \frac{2^{\frac{n+4}{2}}}{D B(\beta, \beta)} \left[ \left( \frac{n}{2}! \right)^{2} \Phi_{2\beta}^{*}(-1, n+1, \beta) + \alpha^{2} \left( \frac{n+2}{2}! \right)^{2} \Phi_{2\beta}^{*}(-1, n+3, \beta) \right]$$

## In particular:

i. 
$$E(Z) = {-4\alpha\psi^{(1)}(\beta)}/D = \mu(say)$$
.

ii. 
$$E(Z^2) = \frac{2}{D} \left[ 2 \psi^{(1)}(\beta) + \alpha^2 \{ 6(\psi^{(1)}(\beta))^2 + \psi^{(3)}(\beta) \} \right].$$

iii. 
$$\operatorname{Var}(Z) = \frac{2}{D^2} \Big[ 2\alpha^2 (3D - 4)(\psi^{(1)}(\beta))^2 + \alpha^2 D\psi^{(3)}(\beta) + 2D\psi^{(1)}(\beta) \Big]$$

**Remark 2.2:** The parameter  $\alpha$  can be expressed in terms of mean ( $\mu$ ) as

$$\alpha = \frac{-1 \pm \sqrt{1 - \frac{\mu^2}{\psi^{(1)}(\beta)}}}{\mu} \ \forall \beta > 1$$

**Result 2.2.** If  $Z \sim ASGLG(\alpha, \beta)$ , then the moment generating function (mgf) is given by

$$M_{Z}(t) = \frac{1}{DB(\beta,\beta)} \left[ \frac{\alpha^{2} I_{1}(t)}{(\beta+t)^{3}} + \frac{\alpha^{2} I_{1}(-t)}{(\beta-t)^{3}} - \frac{\alpha I_{2}(t)}{(\beta+t)^{2}} + \frac{\alpha I_{2}(-t)}{(\beta-t)^{2}} + \frac{I_{3}(-t)}{\beta-t} + \frac{I_{3}(t)}{\beta+t} \right].$$

Where,  $I_1(t) = {}_4F_3(2\beta, t + \beta, t + \beta, t + \beta; t + \beta + 1, t + \beta + 1, t + \beta + 1; -1)$ ,

$$I_2(t) = {}_3F_2(2\beta, t + \beta, t + \beta; t + \beta + 1, t + \beta + 1; -1)$$

$$I_3(t) = {}_{2}F_1(2\beta, \beta - t; -t + \beta + 1; -1)$$

**Remark 2.3.** An alternative expression for the mgf of ASGLG( $\alpha, \beta$ ) can be obtained in terms of gamma function and its derivatives as

$$\frac{1}{\Gamma(\beta)^{2} \left[1 + \alpha^{2} \psi^{(1)}(\beta)\right]} \left[\Gamma(\beta + t) \Gamma(\beta - t) - \alpha \left\{\Gamma^{\prime}(\beta + t) \Gamma(\beta - t) - \Gamma(\beta + t) \Gamma^{\prime}(\beta - t)\right\} + \frac{\alpha^{2}}{2} \left\{\Gamma^{\prime\prime}(\beta + t) \Gamma(\beta - t) + \Gamma^{\prime}(\beta + t) \Gamma^{\prime\prime}(\beta - t)\right\} \right]$$
(8)

#### Reference

- [1] Azzalini, A (1985) A class of distributions which includes the normal ones. Scand. J. Stat. 12: 171 178.
- [2] Huang, W. J., Chen, Y. H. (2007) Generalized skew Cauchy distribution, Statist Probab Lett. 77:1137 1147.
- [3] Chakraborty, S., Hazarika, P. J. (2011) A survey of the theoretical developments in univariate skew normal distributions. Assam Statist. Rev. 25(1): 41 63.
- [4] Elal-Olivero, D. (2010) Alpha-skew-normal distribution, Proyecciones, **29**: 224-240.
- [5] Handam, A. H. (2012) A note on generalized alpha-skew-normal distribution International Journal of Pure and Applied Mathematics Volume 74 No. 4 2012, 491-496
- [6] Harandi, S. S., Alamatsaz, M. H. (2013) Alpha-skew-Laplace distribution, Statist Probab Lett. 83(3):774 782.
- [7] Hazarika, P. J., Chakraborty, S. (2014) Alpha Skew Logistic Distribution. IOSR Journal of Mathematics. 10(4) Ver. VI: 36-46
- [8] Balakrishnan, N. (1992) Handbook of Logistic Distribution. Marcel Dekker. New York

- [9] Johnson, N. L., Kotz, S. and Balakrishnan, N. (2004) Continuous Univariate Distributions (Volume 2). A Wiley-Interscience Publication. John Wiley & Sons Inc. New York
- [10] Gradshteyn, I. S., Ryzhik, I. M. (2007) Table of Integrals, Series, and Products, (7<sup>th</sup> edn.). Academic Press, USA.
- [11] Pogány, T. K., Nadarajah, S. (2013) Moments of generalized logistic random Variables, <a href="http://dx.doi.org/10.1080/10652469.2013.838758">http://dx.doi.org/10.1080/10652469.2013.838758</a>
- [12] Goyal, S. P., Laddha R. K. (1997) On the generalized Riemann–Zeta functions and the generalized Lambert transform. Gan. Ita Sandesh.; 11: 97–108.

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