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# **Short Communication**

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# Maximum Likelihood Estimation in an Alpha-Power

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#### **Abstract**

In this paper a new kind of alpha-power transformed family of distributions (APTA-F for short) is introduced. We discuss the maximum likelihood method of estimating the unknown parameters in a sub-model of this new family. The simulation study performed indicates the method of maximum likelihood is adequate in estimating the unknown parameters in the new family. The applications indicate the new family can be used to model real life data in various disciplines. Finally, we propose obtaining some properties of this new family in the conclusions section of the paper.

Keywords: Maximum likelihood estimation; Monte carlo simulation; Alpha-power transformation; Guinea Big data

#### Introduction

Mahdavi and Kundu [1] proposed the celebrated APT- ${\it F}$  family of distributions with CDF

$$G(x;\alpha,\xi) = \frac{\alpha^F(x;\xi)-1}{\alpha-1}$$

where  $1 \neq \alpha > 0, x \in \mathbb{R}$  and  $\xi$  is a vector of parameters all of whose entries are positive. By modifying the above CDF, the Zubair-G family of distributions appeared in [2] with the following CDF

$$F(x;\alpha,\xi) = \frac{e^{\alpha G(x;\xi)^2} - 1}{e^{\alpha} - 1}$$

where  $\alpha > 0, x \in \mathbb{R}$  and  $\zeta$  is a vector of parameters all of whose entries are positive. Subsequently in [3] we introduced the Ampadu-G family of distributions by modifying the above CDF, to give the following CDF

$$F(x;\lambda,\xi) = \frac{1 - e^{-\lambda G(x;\xi)^2}}{1 - e^{-\lambda}}$$

where  $\lambda>0, x\in\mathbb{R}$ , and  $\xi$  is a vector of parameters all of whose entries are positive. Consequently, we introduced new variants of the APT-F family of distributions, which are recorded in the table below (Table 1).

Table 1: Variants of the Alpha Power Transformed Family of Distributions

Year	Name of Distribution	Author
2019	Ampadu APT -G	Ampadu [4]
2019	$\left(\frac{1}{e}\right)^{\alpha}PT-G$	Ampadu [4]
2019	$\left(\frac{1}{e}\right)^{a} PT - F$ Within Quantile Distribution (New)	Ampadu and to appear in [5]
2019	APTA-F (Current Manuscript)	Ampadu

The rest of this paper is organized as follows. In Section 2 we introduce and illustrate the new family. In Section 3 we derive the quantile function of the APTA-F family of distributions. In Section 4, we discuss approximate random sampling from the APTA-F family of distributions, and maximum likelihood stimation in Section 5. Section 6 presents simulation results associated with the APTA-F family of distributions, assuming F follows the Weibull distribution as defined in Section 2. Section 7 discusses usefulness of the APTA-F family of distributions in fitting real-life data. Section 8 is devoted to the conclusions.

## The New Family

Definition 2.1. Let  $x \in \mathbb{R}$ ,  $\alpha \in \mathbb{R}$ , and  $F(x;\xi)$  be a baseline CDF all of whose parameters are in the vector  $\xi$ . We say a random variable X follows the alpha-power transformed distribution of the Ampadu type (APTA-F for short) if the CDF is given by



$$\frac{F(x;\xi)e^{-\alpha F(x;\xi)}}{e^{-\alpha}}$$

If the baseline distribution is given by the Weibull distribution with  $\ensuremath{\mathsf{CDF}}$ 

$$1-e^{-\left(\frac{x}{b}\right)^a}$$

for x,a,b>0 , then we have the following from the above definition

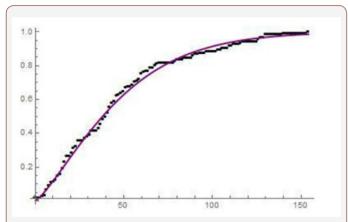
Proposition 2.2. The CDF of the APTA-Weibull distribution is given by

$$\left(1-e^{-\left(\frac{x}{b}\right)^{a}}\right)e^{\alpha e^{-\left(\frac{x}{b}\right)^{a}}}$$

Where x, a, b > 0 and  $\alpha \in \mathbb{R}$ 

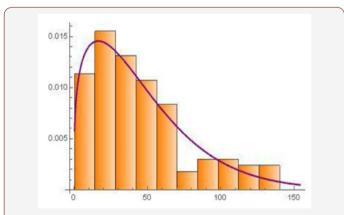
Notation 2.3. If a random variable V follows the APTA-Weibull distribution, we write

$$V \sim APTAW(a,b,\alpha)$$
 (Figure 1)



**Figure 1:** The CDF of APTAW (1.28352800, 48.41833872, -0.07850413) fitted to the empirical distribution of the data on patients with breast cancer, Section 5.2 [6].

By differentiating the CDF of the APTA-F distribution we have the following



**Figure 2:** The PDF of APTAW (1.28352800, 48.41833872, -0.07850413) fitted to the histogram of the data on patients with breast cancer, Section 5.2 [6].

Proposition 2.4. The PDF of the alpha-power transformed distribution of the Ampadu type is given by

$$e^{\alpha-\alpha F(x,\xi)} (1-\alpha F(x,\xi)) f(x,\xi)$$

where  $x \in \mathbb{R}, \alpha \in \mathbb{R}$  and  $F(x;\xi)$  and  $f(x;\xi)$  are the CDF and PDF, respectively, of some basline distribution all of whose parameters are in the vector  $\xi$  (Figure 2)

# The Quantile

**Theorem 3.1.** Let  $Q_F(\cdot) = F^{-1}(\cdot)$  denote the quantile of some baseline distribution with CDF F and PDF f,  $\alpha \in \mathbb{R}$  and 0 < u < 1. The quantile function of the alpha-power transformed distribution of the ampadu type is given by

$$Q_F \left[ - \frac{W \left( - lpha e^{-lpha} u 
ight)}{lpha} \right]$$

Where  $W(z) \coloneqq \Pr{oductLog(z)}$  gives the principal solution for w in  $z = we^w$ 

Proof. Let 0 < u < 1. Since  $Q_F(\cdot) = F^{-1}(\cdot)$  the quantile function can be obtained by solving the following equation for y

$$u = \frac{F(y)e^{-\alpha F(y)}}{e^{-\alpha}}$$

# **Approximate Random Number Generation**

It is well known that the principal solution for w in  $z=we^w$ , denoted W(z):=  $\Pr{oductLog(z)}$  admit the following Taylor series expansion [7]

$$\sum_{n=1}^{\infty} \frac{\left(-n\right)^{n-1}}{n!} z^n = z - z^2 + \frac{3}{2} z^3 + \frac{8}{3} z^4 + \frac{125}{24} z^5 - \dots$$

Using the first term of this series to approximate  $W(z) := \Pr{oductLog(z)}$ , gives us a way to approximate the random sample from the APTA-F family of distributions.

In particular if  $u \sim U(0,1)$ , that is, u is a uniform random variable, then an (approximate) random sample from the APTA-F family of distributions can be obtained via

$$X = Q_F \left( e^{-\alpha} u \right)$$

where  $Q_F(\cdot) = F^{-1}(\cdot)$  denote the quantile of some baseline distribution with CDF F and PDF f,  $\alpha \in \mathbb{R}$ 

#### **Parameter Estimation**

In this section, we obtain the maximum likelihood estimators (MLEs) for the parameters of the APTA-F family of distributions. For this, let  $X_1, X_2, .... X_n$  be a random sample of size n from the APTA-F family of distributions. The likelihood function from Proposition 2.4 is given by

$$L = \prod_{i=1}^{n} \left\{ e^{-\alpha - \alpha F(x_i;\xi)} \left( 1 - \alpha F(x_i;\xi) \right) f(x_i;\xi) \right\}$$

From the above the log-likelihood function is given by

$$\ln L = \sum_{i=1}^{n} \alpha \left(1 - F\left(x_{i}; \xi\right)\right) + \sum_{i=1}^{n} \ln \left(1 - \alpha F\left(x_{i}; \xi\right)\right) + \sum_{i=1}^{n} \ln f\left(x_{i}; \xi\right)$$

The MLE's of  $\xi$  and can be obtained by maximizing the equation immediately above.

The derivatives of the equation immediately above with respect to the unknown parameters, are given as follows

$$\ln L = \sum_{i=1}^{n} \alpha \left( 1 - F\left(x_{i}; \xi\right) \right) + \sum_{i=1}^{n} \ln \left( 1 - \alpha F\left(x_{i}; \xi\right) \right) + \sum_{i=1}^{n} \ln f\left(x_{i}; \xi\right)$$

$$\frac{\partial \ln L}{\partial \alpha} = \sum_{i=1}^{n} \ln \left( 1 - F\left(x_{i}; \xi\right) \right) + \sum_{i=1}^{n} \frac{F\left(x_{i}; \xi\right)}{\alpha F\left(x_{i}; \xi\right) - 1}$$

$$\frac{\partial \ln L}{\partial \xi} = -\sum_{i=1}^{n} \alpha \frac{\partial F(x_i; \xi)}{\partial \xi} + \sum_{i=1}^{n} \frac{\alpha \frac{\partial F(x_i; \xi)}{\partial \xi}}{\alpha F(x_i; \xi) - 1} + \sum_{i=1}^{n} \frac{\partial f(x_i; \xi)}{\partial \xi}$$

Now solving the system below for \_ and \_ gives the maximum likelihood estimators,  $\hat{\alpha}$  and  $\hat{\xi}$  of the unknown parameters:

$$\frac{\partial \ln L}{\partial \alpha} = 0$$
 and  $\frac{\partial \ln L}{\partial \xi} = 0$ 

#### **Simulation Studies**

In this section, a Monte Carlo simulation study is carried out to assess the performance of the estimation method, when the baseline distribution is defined as in Section 2.

#### **Simulation Study One**

Approximate samples of sizes 200, 350, 500, and 700, are drawn from the APTA-Weibull distribution. The approximate samples have been drawn for  $(a,b,\alpha)$  = (1.3, 48.4, 0) using

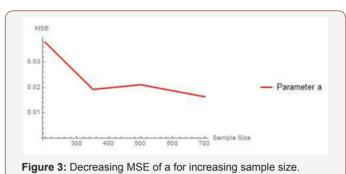
$$X = b\left(-\log\left(1 - e^{-\alpha}u\right)\right)^{1/a}$$

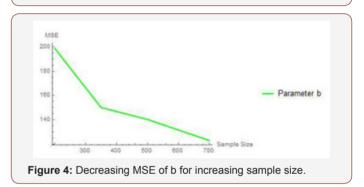
Table 2: Result of Simulation Study.

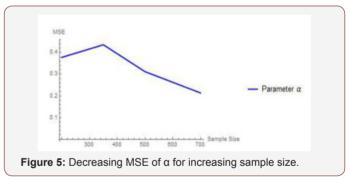
Parameter a				
Sample Size	Average Estimate	MSE		
200	1.276885	0.037847		
350	1.269485	0.019242		
500	1.261209	0.021077		
700	1.268349	0.016317		
	Parameter b			
Sample Size	Average Estimate	MSE		
200	48.55782	198.7113		
350	48.3844	150.1065		
500	48.73268	140.3203		
700	48.10531	123.0901		
	Parameter α			
Sample Size	Average Estimate	MSE		
200	-0.10479	0.377691		
350	-0.16417	0.436073		
500	-0.08719	0.311987		
700	-0.06227	0.213638		

and the maximum likelihood estimators for the parameters a, b, and  $\alpha$  are obtained. The procedure has been repeated 200 times and the mean and mean square error for the estimates are computed, and the results are summarized in Table 2 below.

From Table 2 above, we find that the simulated estimates are close to the true values of the parameters and hence the estimation method is adequate. We have also observed that the estimated mean square errors (MSEs) consistently decrease with increasing sample size as seen in the Figures below.







#### Simulation study two

Approximate samples of sizes 400, 550, 700, and 900, are drawn from the APTA-Weibull distribution. The approximate samples have been drawn for  $(a,b,\alpha) = (1.3, 1.3, 0)$  using

$$X = b\left(-\log\left(1 - e^{-\alpha}u\right)\right)^{1/a}$$

and the maximum likelihood estimators for the parameters a, b, and  $\_$  are obtained. The procedure has been repeated 400 times and the bias and root mean square error for the estimates are computed, and the results are summarized in Table 3 below.

From Table 3 above, we find that the estimated root mean square errors (RMSEs) consistently decrease with increasing sample size as seen in the Figures below. We also find that the bias

is consistently around zero, hence estimating the parameters in the distribution via the method of maximum likelihood is adequate.

Table 3: Result of Simulation Study.

Parameter a			
Sample Size	Bias	RMSE	
400	-0.01948298	0.1563376	
550	-0.02780227	0.1355823	
700	-0.01273009	0.1193164	
900	-0.01177412	0.1100411	
Parameter b			
Sample Size	Bias	RMSE	
400	0.02225872	0.3380576	
550	-0.004792135	0.3157712	
700	0.007635633	0.2872791	
900	0.009964405	0.2334487	
Parameter α			
Sample Size	Bias	RMSE	
400	-0.1131752	0.5872724	
550	-0.07089181	0.5248752	
700	-0.09093898	0.505457	
900	-0.03949046	0.430892	

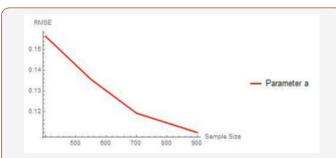
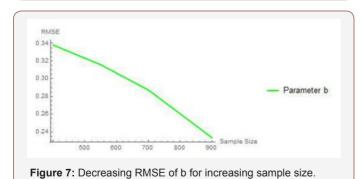


Figure 6: Decreasing RMSE of a for increasing sample size.



Parameter  $\alpha$ 0.50

0.45

0.50

0.45

Sample Size

Figure 8: Decreasing RMSE of  $\alpha$  for increasing sample size.

# **Application**

In this section, we illustrate the usefulness of the APTA-F family of distributions in modelling real-life data. We compare the fit of the APTA-Weibull distribution, with the proposed Weibull distribution appearing in [5] to the dataset on the survival times (in days) of 72 guinea pigs infected with virulent tubercle bacilli. The dataset can be found in [8]. The estimates of the unknown parameters in both distributions are obtained by the maximum likelihood method using the R language software. The measures of goodness of fit considered included Akaike information criterion (AIC), consistent Akaike information criterion (CAIC), Bayesian information criterion (BIC) and Hannan-Quinn information criterion (HQIC) statistics and they are defined as follows:

$$AIC = 2l + 2k$$

$$BIC = 2l + k \log(n)$$

$$CAIC = AIC + \frac{2k(k+1)}{n-k-1}$$

$$HQIC = 2\log[\log(n)(k-2l)]$$

where k is the number of parameters in the statistical model, n is the sample size, and  $l(\cdot)$  is the maximized value of the log-likelihood function under the considered model. The proposedWeibull distribution appearing in [5], which we denote, Proposed Weibull  $(a,b,\alpha)$  has CDF given by

$$\frac{1-e^{-a\left(1-e^{-\left(\frac{x}{b}\right)^{a}}\right)}}{1-e^{-a}}$$

Where 
$$\alpha \neq \frac{1}{e}, \alpha > \frac{1}{e}, a, b > 0$$
, and  $x > 0$ 

Since the Proposed Weibull  $(a,b,\alpha)$  distribution has the smallest AIC, BIC, CAIC, and HQIC values compared to the other distribution in Table 4, it can be considered a better fit to the survival times (in days) of 72 guinea pigs infected with virulent tubercle bacilli The asymptotic variance-covariance matrix of the MLEs under the  $APTAW(a,b,\alpha)$ , and the Proposed Weibull  $(a,b,\alpha)$  distributions, respectively, are given by

**Table 4:** Estimated Parameters for the dataset on the survival times (in days) of 72 guinea pigs infected with virulent tubercle bacilli.

Model	Parameter Estimate	Standard Error
APTAW (a, b, α)	$(\hat{a}, \hat{b}, \hat{\alpha})$ = (2.0665258, 265.3997306, 0.7428645)	(0.1801981, 37.1849579, 0.2413815)
Proposed Weibull (a, b, α)	$(\hat{a}, \hat{b}, \hat{\alpha}) = (2.208316, 310.356991, 3.239273)$	(0.1968281, 70.0390762,1.7622610)

$$\begin{bmatrix} 3.247136e - 02 & -0.6552024 & -3.845612e - 05 \\ -6.552024e - 01 & 1382.7210969 & 7.889494e + 00 \\ -3.845612e - 05 & 7.8894941 & 5.826503e - 02 \end{bmatrix}$$
$$\begin{bmatrix} 0.03874132 & -0.7365994 & 0.03693494 \\ -0.73659941 & 4905.4721882 & 116.48517435 \\ 0.03693494 & 116.4851744 & 3.10556397 \end{bmatrix}$$

Hence the approximate 95% confidence intervals for the parameters under the  $APTAW(a,b,\alpha)$  and the Proposed Weibull distributions, respectively, are given in Tables 5-7.

Table 5: Criteria for Comparison.

Model	AIC	BIC	CAIC	HQIC
AP T AW (a, b, α)	857.8961	864.7261	858.2491	16.40831
Proposed Weibull (a, b, α)	856.7445	863.5745	857.0974	16.40561

Table 6: APTAW (a, b, α) distribution.

Parameter	CI
a	(2.055226, 2.077825)
b	(263.0680, 267.7315)
α	(0.7277283,0.7580008)

**Table 7:** Proposed Weibull (a, b,  $\alpha$ ) distribution.

Parameter	CI
а	(2.195974 2.220659)
b	(305.9651, 314.7489)
α	(3.128767 3.349778)

## **Concluding Remarks**

In this paper a new kind of alpha power transformed family of distributions is introduced, and members of this family are shown to be effective in fitting real life data. As a further development we propose obtaining some mathematical and statistical properties of the APTA-*F* family of distributions.

#### Acknowledgement

None.

# **Conflict of Interest**

No conflict of interest.

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