# CHARACTERIZATIONS OF LIFE DISTRIBUTIONS FROM PERCENTILE RESIDUAL LIFETIMES

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## Summary

An  $\alpha$ -percentile residual life function does not uniquely determine a life distribution; however, a continuous life distribution can be uniquely determined by its  $\alpha$ -percentile and  $\beta$ -percentile residual life functions if  $\alpha$  and  $\beta$  satisfy a certain condition. Two characterizations in terms of percentile residual lifetimes are given for the Beta  $(1, \theta, K)$ , Exponential  $(\lambda)$  and Pareto  $(\theta, K)$  family of distributions.

#### 1. Introduction

Let F(t) be a life distribution. The mean and  $\alpha$ -percentile residual life functions at time t are respectively the mean and  $100\alpha$  percentile residual life given survival up to time t (mathematical definitions are given in Section 2). It is well-known that if F has finite mean, then it is uniquely determined by its mean residual life function  $m_r(t)$ —see, for example, Meilijson [6], Swartz [9], Laurent [5], Galambos and Kotz [2], and Hall and Wellner [3]. In contrast, it is shown in Arnold and Brockett [1] and Joe and Proschan [4] that there can be infinitely many life distributions with the same  $\alpha$ -percentile residual life function  $q_{a,p}(t)$ . Hence it is impossible to characterize life distributions in terms of a single percentile residual life function. In this paper, we show that a continuous life distribution F is uniquely determined by  $q_{\alpha,F}$  and  $q_{\theta,F}$  if  $\alpha$  and  $\beta$  satisfy a certain condition. The Beta  $(1,\theta,K)$ , Exponential ( $\lambda$ ) and Pareto ( $\theta$ , K) (of the second kind) distributions are characterized as the only absolutely continuous distributions with (i) linear  $\alpha$ -percentile residual lifetimes for an interval of  $\alpha$ 's and (ii) the mean residual life function equivalent to an  $\alpha$ -percentile residual life function for some  $0 < \alpha < 1$ . Characterizations in terms of linear mean residual life are given in Hall and Wellner [3] and Morrison [7].

Key words and phrases: Residual life, percentile, beta, exponential, Pareto.

# 2. Definitions

Let F be a life distribution, that is, a distribution such that  $F(0^-) = 0$ , and let  $\overline{F}$  be the survival function:  $\overline{F}(t) = 1 - F(t)$ , t > 0. Let  $F^{-1}$  be the left continuous inverse of F defined by

$$F^{-1}(u) = \inf \{x: F(x) \ge u\}, \quad 0 \le u \le 1.$$

Then  $\bar{F}^{-1}(u)=F^{-1}(1-u)$  is the right continuous inverse of  $\bar{F}$ . Let  $T_F=\sup\{x\colon F(x)<1\}=F^{-1}(1)$  be the right-hand endpoint of support and for  $0\le t\le T_F$ , let

$$egin{align*} r_F(t) &= f(t)/ar{F}(t) \;, \ &ar{F}_t(x) = 1 - F_t(x) = ar{F}(t+x)/ar{F}(t) \;, \ &m_F(t) = \int_0^\infty ar{F}_t(x) dx = \int_t^\infty ar{F}(x) dx/ar{F}(t) \;, \quad & ext{and} \ &q_{\sigma,F}(t) = F_t^{-1}(\alpha) = F^{-1}(1 - ar{\alpha}ar{F}(t)) - t = ar{F}^{-1}(ar{\alpha}ar{F}(t)) - t \;. \end{split}$$

be the failure rate, the conditional survival function, the mean residual life function and the  $\alpha$ -percentile residual life function, respectively. Whenever possible, we will suppress the subscript F and use T, r, m,  $q_{\alpha}$  in place of  $T_F$ ,  $r_F$ ,  $m_F$ ,  $q_{\alpha,F}$  respectively.

## 3. Uniqueness in terms of two percentile residual lifetimes

In Theorem 1 below, a functional relationship is obtained for two continuous life distributions F and G with the same  $\alpha$ -percentile residual life function. Using this result, it is shown that a continuous life distribution is uniquely determined by its  $\alpha$ - and  $\beta$ -percentile residual life functions if  $\alpha$  and  $\beta$  satisfy the condition in Theorem 2.

THEOREM 1. Let F and G be continuous life distributions and let  $0 < \alpha < 1$ . Then  $q_{\alpha,F} \equiv q_{\alpha,G}$  if and only if there exists a periodic function  $\Delta(x)$  with period  $-\log(1-\alpha)$ , such that

(1) 
$$\overline{F}(x) = \Delta(-\log \overline{G}(x)) \cdot \overline{G}(x), \quad x \ge 0.$$

PROOF. (Necessity). Suppose that  $q_{\alpha,F} \equiv q_{\alpha,G}$ . Then  $\overline{F}^{-1}(\overline{\alpha}\overline{F}(t)) = \overline{G}^{-1}(\overline{\alpha}\overline{G}(t)), t \geq 0$ , which implies

(2) 
$$\bar{\alpha}\bar{F}(t) = \bar{F}\bar{G}^{-1}(\bar{\alpha}\bar{G}(t)), \qquad t \ge 0$$

since F and G are continuous. The hypotheses imply that F is constant on any interval where G is constant and vice-versa so that (2)

implies

(3) 
$$\bar{\alpha}\bar{F}\bar{G}^{-1}(u) = \bar{F}\bar{G}^{-1}(\bar{\alpha}u), \quad 0 < u < 1.$$

The function

$$\Delta(x) = \bar{F}\bar{G}(e^{-x})e^x, \quad x \ge 0$$

satisfies (1), and  $\Delta(x) = \Delta(x - \log \bar{\alpha})$ ,  $x \ge 0$ , if and only if

$$\bar{F}\bar{G}^{-1}(e^{-x}) = \bar{F}\bar{G}^{-1}(\bar{\alpha}e^{-x})/\bar{\alpha}$$
,  $x \ge 0$ 

which is equivalent to (3).

(Sufficiency). Suppose that (1) holds for a periodic function with period  $-\log(1-\alpha)$ . Then

$$ar{F}ar{G}^{-1}(ar{a}ar{G}(t)) = \Delta(-\log ar{a}ar{G}(t)) \cdot ar{a}ar{G}(t)$$

$$= \Delta(-\log ar{G}(t)) \cdot ar{a}ar{G}(t) = ar{a}ar{F}(t) , \qquad t \ge 0 ,$$

and equation (2) holds. For fixed  $t \ge 0$ ,  $\bar{F}^{-1}(\bar{a}\bar{F}(t)) = \bar{F}^{-1}(\bar{F}(y))$ , where  $y = \bar{G}^{-1}(\bar{a}\bar{G}(t))$ . Let  $I = [t_1, t_2]$  be the interval of u's for which  $\bar{F}(u) = \bar{F}(y)$ . Then  $\bar{F}^{-1}(\bar{F}(y)) = F^{-1}(F(y)) = t_1$ . (2) implies that  $\bar{F}(x)$  is constant on an interval  $I \Longleftrightarrow \bar{G}^{-1}(\bar{a}\bar{G}(x))$  is constant on  $I \Longleftrightarrow \bar{G}(x)$  is constant on I. Therefore,  $\bar{G}(t_1) = \bar{G}(y) = \bar{G}(t_2) = \bar{a}\bar{G}(t)$  and  $\bar{G}^{-1}(\bar{a}\bar{G}(t)) = t_1 = y$ . Hence  $\bar{F}^{-1}(\bar{a}\bar{F}(t)) = \bar{G}^{-1}(\bar{a}\bar{G}(t))$  and  $q_{a,F}(t) = q_{a,G}(t)$ .

THEOREM 2. Let F and G be continuous life distributions. If  $q_{\alpha,F} \equiv q_{\alpha,G}$  and  $q_{\beta,F} \equiv q_{\beta,G}$  for  $\alpha$  and  $\beta$  in (0,1) such that  $\log (1-\alpha)/\log (1-\beta)$  is irrational, then  $F \equiv G$ .

PROOF. Let  $\Delta(x)$  be defined as in (4). By Theorem 1,  $\Delta$  is periodic with periods  $-\log(1-\alpha)$  and  $-\log(1-\beta)$ . Since  $\log(1-\alpha)/\log(1-\beta)$  is irrational,  $\Delta$  is periodic with a dense set of periods and hence is constant (Semadeni [8]). The hypotheses of the theorem imply that F(0) = G(0) = 0 and therefore, from (1) of Theorem 1,  $F \equiv G$ .

Remarks. 1. Arnold and Brockett [1] have a proof of Theorem 2 assuming that F and G are strictly incresing.

- 2. A corollary of Theorem 2 is: If F and G are continuous life distributions such that  $q_{\beta,F} \equiv q_{\beta,G}$  for all  $\beta \in I$ , where I is a subinterval of (0,1), then  $F \equiv G$ . This generalizes Theorem 13 of Joe and Proschan [4].
- 3. Theorem 2 generalizes the following result in Joe and Proschan [4]: If  $q_{\alpha,F}$  and  $q_{\beta,F}$  are constant functions and  $\log(1-\alpha)/\log(1-\beta)$  is irrational, then F is an exponential distribution.

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# 4. Characterization for Beta, Exponential and Pareto distributions

In Remark 3 following Theorem 2, a characterization is given in terms of 2 constant percentile residual lifetimes. The next simplest form of a percentile residual lifetime is a linear one, and for a characterization we will need to consider a continuum of  $\alpha$ 's. The characterization is for the Beta  $(1, \theta, K)$ , Exponential  $(\lambda)$  and Pareto  $(\theta, K)$  (of the second kind) distributions. The survival functions for these 3 families together with the corresponding failure rate, mean residual life and  $\alpha$ -percentile residual life functions are stated next before the characterization result. Note that a Beta  $(1, \theta, K)$  distribution is the distribution of K times a Beta  $(1, \theta)$  random variable.

(i) Beta 
$$(1, \theta, K)$$
,  $0 \le t \le K$ ,  $\theta > 0$ ,  $K > 0$ :
$$\bar{F}(t) = (1 - t/K)^{\theta}, \qquad r(t) = \theta/(K - t),$$

$$m(t) = (K - t)/(\theta + 1), \quad \text{and} \quad q_{\sigma}(t) = (1 - \bar{q}^{1/\theta})(K - t).$$

(ii) Exponential ( $\lambda$ ),  $t \ge 0$ ,  $\lambda > 0$ :

$$ar{F}(t) = e^{-\lambda t}$$
,  $r(t) = \lambda$ ,  $m(t) = 1/\lambda$ , and  $q_{\alpha}(t) = (-\log \overline{\alpha})/\lambda$ .

(iii) Pareto  $(\theta, K)$ ,  $t \ge 0$ ,  $\theta > 0$ , K > 0:

$$ar{F}(t) = (1+t/K)^{-\theta}$$
,  $r(t) = \theta/(K+t)$ ,  $m(t) = (K+t)/(\theta-1)$ ,  $(\theta > 1)$ , and  $q_a(t) = (\overline{\alpha}^{-1/\theta}-1)/(K+t)$ ,  $m(t)$  does not exist for  $0 < \theta \le 1$ .

THEOREM 3. Suppose that F is absolutely continuous and that  $q_{\alpha}(t)$  is linear in t for all  $\alpha \in I$  for some subinterval I of (0,1). Then F is either a Beta  $(1, \theta, K)$ , Exponential  $(\lambda)$  or Pareto  $(\theta, K)$  distribution.

PROOF. Suppose that for  $\alpha \in I$ ,  $q_{\alpha}(t) = a_{\alpha} + b_{\alpha}t$ ,  $0 \le t < T$ , where  $a_{\alpha} > 0$  and  $b_{\alpha} > -1$ . Note that  $q_{\alpha}(t) + t = \bar{F}^{-1}(\bar{\alpha}\bar{F}(t))$  is increasing in t. Since F is continuous, then

(5) 
$$\bar{\alpha}\bar{F}(t) = \bar{F}(q_{\alpha}(t)+t) = \bar{F}(a_{\alpha}+(b_{\alpha}+1)t), \quad 0 \leq t < T.$$

It follows that  $a_{\alpha}$  and  $b_{\alpha}$  are differentiable in  $\alpha$ . Let the two derivatives be denoted by  $a'_{\alpha}$  and  $b'_{\alpha}$ . Differentiation of (5) with respect to  $\alpha$  and with respect to t yield

$$\bar{F}(t) = f(a_{\alpha} + (b_{\alpha} + 1)t) \cdot (a'_{\alpha} + b'_{\alpha}t)$$
 and  $\bar{a}f(t) = f(a_{\alpha} + (b+1)t) \cdot (b_{\alpha} + 1)$ .

The ratio of these two equations leads to

$$1/r(t) = \overline{\alpha}(\alpha'_{\alpha} + b'_{\alpha}t)/(b_{\alpha} + 1)$$
,  $0 \le t < T$ ,  $\alpha \in I$ .

Thus  $\bar{a}a'_a/(b_\alpha+1)$  and  $\bar{a}b'_a/(b_\alpha+1)$  are constants, say,  $c_1$  and  $c_2$ , with  $c_1>0$ . Since a failure rate function uniquely determines a survival function, the only possibilities for F are:

- (i) F is Beta  $(1, \theta, K)$  if  $c_1 = K/\theta$ ,  $c_2 = -1/\theta$ ,
- (ii) F is Exponential ( $\lambda$ ) if  $c_1 = 1/\lambda$ ,  $c_2 = 0$ , and
- (iii) F is Pareto  $(\theta, K)$  if  $c_1 = K/\theta$ ,  $c_2 = 1/\theta$ .

Remark. We conjecture that the conclusion of Theorem 3 is still valid if the hypothesis of absolute continuity is replaced by F(0)=0, F is nondegenerate and F has no mass at  $T=F^{-1}(1)$ .

Finally, we characterize all absolutely continuous distributions for which m(t) is equivalent to  $q_a(t)$  for some  $0 < \alpha < 1$ . We first note the following.

- (i) Within the Beta  $(1, \theta, K)$  family,  $m(t) \equiv q_a(t)$  if  $\alpha = 1 (1 + 1/\theta)^{-\theta} = \alpha_1(\theta)$ .  $\alpha_1(\theta)$  is strictly increasing in  $\theta > 0$  with  $\lim_{\theta \to 0} \alpha_1(\theta) = 0$  and  $\lim_{\theta \to 0} \alpha_1(\theta) = 1 e^{-1}$ .
- (ii) Within the Exponential ( $\lambda$ ) family,  $m(t) \equiv q_{\alpha}(t)$  if  $\alpha = 1 e^{-1}$ .
- (iii) Within the Pareto  $(\theta, K)$  family,  $m(t) \equiv q_{\alpha}(t)$  if  $\alpha = 1 (1 1/\theta)^{\theta} = \alpha_{2}(\theta)$ .  $\alpha_{2}(\theta)$  is strictly decreasing in  $\theta > 1$  with  $\lim_{\theta \to \infty} \alpha_{2}(\theta) = 1 e^{-1}$  and  $\lim_{\theta \to \infty} \alpha_{2}(\theta) = 1$ .

THEOREM 4. Suppose that  $m(t) \equiv q_{\alpha}(t)$  for some  $\alpha \in (0, 1)$ . If F is absolutely continuous, then there is a  $\mu \geq 0$   $(\mu = F^{-1}(0))$  such that  $G(t) = F(t-\mu)$  is either a Beta  $(1, \theta, K)$  or an Exponential  $(\lambda)$  or a Pareto  $(\theta, K)$  distribution.

Remark. A Pareto  $(\theta, K)$  distribution shifted by  $\mu = K$  is a Pareto distribution of the first kind with  $\bar{F}(t) = (K/t)^{\theta}$ ,  $t \ge K$ .

PROOF. If F is absolutely continuous, then from the definition of m(t) and the equation

$$|\bar{F}(t)| = \exp\left\{-\int_0^t r(y)dy\right\} = [m(0)/m(t)]\bar{F}(0) \exp\left\{-\int_0^t [m(y)]^{-1}dy\right\},$$

m'(t) exists and

(6) 
$$r(t) = (1 + m'(t))/m(t), \quad 0 \le t < T.$$

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From the hypotheses of the theorem,

$$\bar{\alpha}\bar{F}(t) = \bar{F}(q_a(t)+t) = \bar{F}(m(t)+t)$$
.

Differentiation with respect to t leads to

(7) 
$$r(t) = r(t+m(t)) \cdot (1+m'(t)), \quad 0 \le t < T.$$

Let  $t_1=\sup\{t\colon 1+m'(t)=0,\ 0\leq t< T\}$ . This is well-defined since the continuity of m(t) implies that  $\bar{F}^{-1}(\bar{\alpha}\bar{F}(t))$  is continuous which in turn implies that  $F^{-1}(u)$  is continuous for  $u\geq \alpha$  or 1+m'(t)>0 for  $t>F^{-1}(\alpha)$ . Equating (6) and (7) yields

(8) 
$$r(t+m(t))=1/m(t), t_1 < t < T.$$

Substitution of (6) on the left-hand side of (8) results in

(9) 
$$m'(t+m(t)) = [m(t+m(t))-m(t)]/m(t), t_1 < t < T.$$

Since m(t) is differentiable, then by (8), r(y) is continuous and differentiable for  $y>t_2=t_1+m(t_1)$ . By (6), m'(y) is continuous and differentiable for  $y>t_2$ . Hence both sides of (9) are differentiable,

(10) 
$$m''(t+m(t))(1+m'(t)) = [m'(t+m(t))-m'(t)]/m(t)$$

for  $t_1 < t < T$ , and m''(y) is continuous for  $y < t_2$ . By Taylor's theorem, for  $t < t_2$ ,

(11) 
$$\frac{m(t+m(t))-m(t)}{m(t)}=m'(t+m(t))-\frac{1}{2}m''(s(t))m(t),$$

where s(t) is between t and t+m(t) and can be assumed continuous. Equations (9) and (11) together imply that m''(s(t))=0,  $t>t_2$ , so that m''(y)=0 for  $y>t_2+m(t_2)$ . Hence there is a constant b>-1 such that m'(y)=b for  $y>t_2+m(t_2)$ . By (10), m'(y)=b for  $y>t_2$ . Substitution into (9) yields

(12) 
$$bm(t) = m(t+m(t)) - m(t), \quad t_1 < t < T.$$

Differentiation of (12) results in m'(t)=b,  $t_1 < t < T$ . There is a constant a such that

(13) 
$$m(t) = a + bt$$
,  $t_1 < t < T$ .

We now prove by contradiction that  $F(t_1)=0$ . If  $F(t_1)>0$ , then there are constants  $s_1 < s_2 < t_1$  such that

(14) 
$$1+m'(t)=0 \text{ for } s_2 < t < t_1,$$

and F(s)>0, 1+m'(s)>0 and  $s+m(s)>t_1$  for  $s_1 < s < s_2$ . By (6) and (7),

$$r(s+m(s))=(1+b)/(a+b[s+m(s)])=1/m(s)$$
,  $s_1 < s < s_2$ 

 $\mathbf{or}$ 

(15) 
$$m(s) = a + bs$$
,  $s_1 < s < s_2$ .

(13), (14), (15) together contradict the fact that m(t) is continuous, and therefore  $F(t_1)=0$  and  $t_1=F^{-1}(0)$ .

COROLLARY. If F is absolutely continuous and  $F_t(x)$  is a symmetric distribution for all  $0 \le t < T$ , then F is a Uniform  $[\mu, \mu + K]$  distribution for some  $\mu > 0$ , K > 0.

PROOF. By Theorem 4,  $m(t)=q_{0.5}(t)$  if and only if  $F(t-\mu)$  is a Beta (1,1,K) or Uniform [0,K] distribution. Therefore F is a Uniform  $[\mu,\mu+K]$  distribution; the conditional distribution  $F_t$  is Uniform  $[\max(0,\mu-t),\mu+K-t)$ .

Remark. If the assumption of absolute continuity in Theorem 4 is replaced by F(0)=0 and F is nondegenerate, then the condition  $m(t)\equiv q_a(t)$  for some  $\alpha$  is also satisfied by a subclass of the Geometric family of distributions. If F is Geometric with parameter  $0 , that is, <math>\bar{F}(j)=(1-p)^j$ ,  $j=0,1,\cdots$ , then  $m(t)=q_a(t)=p^{-1}-t+[t]$ ,  $t\geq 0$ , provided  $p^{-1}=n$  is a positive integer and  $1-(1-p)^{n-1}<\alpha \leq 1-(1-p)^n$ . Except for scale and location modifications of these Geometric distributions, we conjecture that there are no other distributions satisfying  $m(t)\equiv q_a(t)$  for some  $\alpha$  under the weakened hypothesis.

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