### A Note on Characterization by Renewal Variable

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**Abstract**: The concept of renewal variable associated with a non-negative random variable X is used to identify the distribution of X as well as its failure rate. Some illustrated examples are given.

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#### Introduction

Let X be a non-negative random variable (often represents the life of a unit in a certain process) with finite mean  $\mu$ , cdf  $F_{x}(.)$  and survival function  $\overline{F}_{x}(.)$ . A new random variable Y with density function  $f_{x}(.)$  can be defined (see, e.g., **Cox**<sup>(1)</sup> as follows:

$$f_Y(x) = \frac{F_X(x)}{\mu} \tag{1.1}$$

The random variable Y has many applications in life length studies (see, e.g., **Scheaffer**<sup>(2)</sup>) as well as in renewal Process (**Zacks**<sup>(3)</sup>). **Gupta**<sup>(4)</sup> has shown that for large values of X, the random variable Y represents the life of the process when an operating component is replaced upon failure by another possessing the same life distribution. **Pakes and Khattree**<sup>(5)</sup> have demonstrated that Y is related to the length biased sampling.

Moreover, several authors have used Y to characterize some probability distributions. Huang and Lin <sup>(6)</sup> have characterized the exponential distribution using a relationship between the  $k^{rR}$  moments of Y and X. Gupta <sup>(2)</sup> has given an explicit formula of the cdf of X in terms of the failure rate of Y.

The main objective of this note is to identify the distribution of X as well as its failure rate function in terms of the mean residual life and failure rate functions of Y.

### 1- The Main Result.

The following Theorem determines the survival function of X as well as its failure rate in terms of the mean residual life function of Y.

Theorem 2.1

Let X be a non-negative continuous random variable with finite mean  $\mu$ , failure rate  $r_X(.)$ , density function  $f_X(.)$  and survival function  $\overline{F_X}(.)$ . Denote by Y, its associated random variable defined by (1.1). Assume that  $g_Y(.)$  is the mean residual life function of Y, then

$$\overline{F}_{X}(x) = \frac{\varepsilon}{g_{Y}^{2}(x)} \left(1 + \vec{g}_{Y}(x)\right) \exp\left(-\int \frac{dx}{g_{Y}(x)}\right), \quad (2.1)$$

For some constant c to be determined from  $F_{x}(0) = 1$ 

$$r_{X}(x) = \frac{1}{g_{Y}(x)} + \frac{2\tilde{g}_{Y}(x)}{g_{Y}(x)} - \frac{\dot{g}_{Y}(x)}{1 + \dot{g}_{Y}(x)}$$
(2.2)

Proof. Using (1.1), the survival function  $\overline{F}_{Y}(x)$ , of Y will be:

$$\bar{F}_{Y}(x) = \mu^{-1} \int_{x}^{\infty} \bar{F}_{X}(z) dz \qquad (2.3)$$

By definition, the mean residual life function of Y (Hall and Wellner  $^{(7)}$ ) is given by:

$$g_Y(\mathbf{x}) = \frac{\int_X^\infty F_Y(z) dz}{F_Y(z)}$$
(2.4)

Using (2.3), we get:  

$$g_{Y}(x) = \frac{\int_{x}^{\infty} \int_{x}^{\infty} F_{X}(t) dt dz}{\int_{x}^{\infty} F_{X}(t) dt} , \text{ i.e.,}$$

$$g_{Y}(x) \int_{x}^{\infty} F_{X}(t) dt = \int_{x}^{\infty} \int_{z}^{\infty} F_{X}(t) dt dz \qquad (2.5)$$

Differentiating (2.5) with respect to x, we get:

$$\frac{\vec{s}_{X}(x)}{\int_{x}^{\infty}\vec{s}_{X}(t)dt} = \frac{1+\vec{g}_{Y}(x)}{g_{Y}(x)}$$
(2.6)

Integrating both sides of (2.6) with respect to x, we get:

$$\int_{\kappa}^{\infty} \overline{F}_{\kappa}(t) dt = \frac{\varepsilon}{g_{\Gamma}(\kappa)} \exp\left(-\int \frac{d\kappa}{g_{\Gamma}(\kappa)}\right)$$
(2.7)

For some constant c to be determined using  $\vec{F}_{x}(0) = 1$ . Differentiating (2.7) with respect to x, we get:

$$\overline{F}_{X}(x) = \frac{c}{g_{Y}^{2}(x)} \left(1 + g_{Y}(x)\right) \exp\left(-\int \frac{dx}{g_{Y}(x)}\right) \quad (2.8)$$

To prove the  $2^{nd}$  result, take the logarithm of both sides of equation(2.8),we get:

$$\ln \overline{F}_{X}(x) = \ln c - 2\ln g_{Y}(x) + \ln(1 + g_{Y}(x)) - \int \frac{dx}{g_{Y}(x)}$$
(2.9)

Differentiating (2.9) with respect to x and recalling that

 $\eta_{X}(\mathbf{x}) = \frac{f_{X}(\mathbf{x})}{F_{X}(\mathbf{x})}$ We get:

$$\eta_{X}(\mathbf{x}) = \frac{1}{g_{Y}(x)} + 2 \frac{\dot{g}_{Y}(x)}{g_{Y}(x)} - \frac{\dot{g}_{Y}(x)}{1 + \dot{g}_{Y}(x)}$$
  
Our proof is complete.

Remark (2.1)

Denote by  $\mathbf{r}_{\mathbf{r}}(.)$  and  $\mathbf{g}_{\mathbf{r}}(.)$ , the failure rate and mean residual life function of the renewal variable respectively and recalling that (**Ruiz and Navarro**<sup>(8</sup>)

$$r_{\mathrm{Y}}(\mathbf{x}) \, \boldsymbol{g}_{\mathrm{Y}}(\mathbf{x}) = 1 + \boldsymbol{g}_{\mathrm{Y}}(\mathbf{x}),$$

then the  $2^{nd}$  part of Theorem (2.1) can be written as follows:

$$r_{X}(x) = 2 r_{Y}(x) - (g_{Y}(x))^{-1} (1 + \frac{\dot{g}_{Y}(x)}{r_{Y}(x)})$$

Examples

(1) If  $g_{\mathbf{F}}(\mathbf{x}) = \mathbf{k} = \text{constant}$ , then equation (2.1) gives:  $\overline{F}_{\mathbf{X}}(\mathbf{x}) = \exp(-\frac{\mathbf{x}}{\mathbf{k}}), \qquad \mathbf{x} > 0$ 

Also equation (2.2) gives  $r_X(x)$  =constant, which is the well known result of the exponential distribution.

(2) If 
$$g_{\mathbb{F}}(\mathbf{x}) = \frac{\omega + \pi}{\vartheta - 2}$$
,  $0 < a \le x$ ,  $\theta > 2$   
Then equations (2.1) and (2.2) give:

 $F(x) = \left(\frac{a+x}{b+x}\right)^{p}$ ,  $0 < a \le x$   $r_{X}(x) = \frac{a}{b+x}$ Which are the well known results for the general Pareto distribution with parameters a, b, and  $\theta$ .

(3) If 
$$g_{Y}(x) = \frac{b-x}{\theta+1}$$
,  $a < x < b$ ,  $\theta > 0$ .  
Then equations (2.1) and (2.2) give:  
 $\overline{F}_{X}(x) = , \quad (\frac{b-x}{b-\alpha})^{\theta}$   
 $r_{X}(x) = \frac{\theta}{b-x}$ 

Which are the well known results for the  $\mathbf{1}^{st}$  type Pearsonian distribution with parameters a,b, and  $\theta > 0$ .

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## Remarks

(1) Similar results for the uniform distribution with parameters a, b can be given. To see this, put  $\theta = 1$  in the last example.

(2) Similar results for the beta distribution with parameters 1,m can be obtained. To this end, put  $\theta =$  m, b =1 and a= 0 in the last example.

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