Deriving an Expression for $P(X(t, t + \tau) = x)$ Under the Pareto/NBD Model

Peter S. Fader www.petefader.com

Bruce G.S. Hardie www.brucehardie.com

Kinshuk Jerath[†]

September 2006 Revised July 2020

1 Introduction

Schmittlein et al. (1987) and Fader and Hardie (2020) derive expressions for P(X(t) = x), where the random variable X(t) denotes the number of transactions observed in the time interval (0, t], as implied by the Pareto/NBD model assumptions. In this note, we derive the corresponding expression for $P(X(t, t + \tau) = x)$, where the random variable $X(t, t + \tau)$ denotes the number of transactions observed in the time interval $(t, t + \tau]$.

In Section 2 we review the assumptions underlying the Pareto/NBD model. In Section 3, we derive an expression for $P(X(t, t + \tau) = x)$ conditional on the unobserved latent characteristics λ and μ . This conditioning is removed in Section 4.

2 Model Assumptions

The Pareto/NBD model is based on the following assumptions:

- i. Customers go through two stages in their "lifetime" with a specific firm: they are "alive" for some period of time, then become permanently inactive (i.e., "die").
- ii. While alive, the number of transactions made by a customer follows a Poisson process with transaction rate λ . This implies that the probability of observing x transactions in the time interval (0, t] is given by

$$P(X(t) = x \mid \lambda) = \frac{(\lambda t)^x e^{-\lambda t}}{x!}, \quad x = 0, 1, 2, \dots$$

It also implies that, assuming the customer is alive through the time interval $(t_a, t_b]$,

$$P(X(t_a, t_b) = x \mid \lambda) = \frac{[\lambda(t_b - t_a)]^x e^{-\lambda(t_b - t_a)}}{x!}, \quad x = 0, 1, 2, \dots$$

 $^{^{\}dagger}$ © 2006, 2020 Peter S. Fader, Bruce G.S. Hardie, and Kinshuk Jerath. This document can be found at <htp://brucehardie.com/notes/013/>.

iii. A customer's unobserved lifetime of length ω (after which he is viewed as being inactive) is exponentially distributed with dropout rate μ :

$$f(\omega \mid \mu) = \mu e^{-\mu \omega} \,.$$

iv. Heterogeneity in transaction rates across customers follows a gamma distribution with shape parameter r and scale parameter α :

$$g(\lambda \,|\, r, \alpha) = \frac{\alpha^r \lambda^{r-1} e^{-\lambda \alpha}}{\Gamma(r)} \,. \tag{1}$$

v. Heterogeneity in dropout rates across customers follows a gamma distribution with shape parameter s and scale parameter β :

$$g(\mu \mid s, \beta) = \frac{\beta^s \mu^{s-1} e^{-\mu\beta}}{\Gamma(s)}.$$
(2)

vi. The transaction rate λ and the dropout rate μ vary independently across customers.

$3 \quad P(X(t,t+\tau)=x) \,\, {\rm Conditional} \,\, {\rm on} \,\, \lambda \,\, {\rm and} \,\, \mu$

Suppose we know an individual's unobserved latent characteristics λ and μ . For x > 0, there are two ways x purchases could have occurred in the interval $(t, t + \tau]$:

i. The individual was alive at t and remained alive through the whole interval; this occurs with probability $e^{-\mu(t+\tau)}$. The probability of the individual making x purchases, given that he was alive during the whole interval, is $(\lambda \tau)^x e^{-\lambda \tau}/x!$. It follows that the probability of remaining alive through the interval $(t, t + \tau)$ and making x purchases is

$$\frac{(\lambda\tau)^x e^{-\lambda\tau} e^{-\mu(t+\tau)}}{x!}.$$
(3)

ii. The individual was alive at t but died at some point ω (< t + τ), making x purchases in the interval (t, ω]. The probability of this occurring is

$$\begin{split} \int_{t}^{t+\tau} & \frac{[\lambda(\omega-t)]^{x}e^{-\lambda(\omega-t)}}{x!} \mu e^{-\mu\omega} \, d\omega \\ &= e^{-\mu t} \lambda^{x} \mu \int_{t}^{t+\tau} \frac{(\omega-t)^{x}e^{-(\lambda+\mu)(\omega-t)}}{x!} \, d\omega \\ &= e^{-\mu t} \lambda^{x} \mu \int_{0}^{\tau} \frac{s^{x}e^{-(\lambda+\mu)s}}{x!} \, ds \\ &= e^{-\mu t} \frac{\lambda^{x} \mu}{(\lambda+\mu)^{x+1}} \int_{0}^{\tau} \frac{(\lambda+\mu)^{x+1}s^{x}e^{-(\lambda+\mu)s}}{x!} \, ds \end{split}$$

which, noting that the integrand is an Erlang-(x + 1) pdf,

$$= e^{-\mu t} \left(\frac{\lambda}{\lambda+\mu}\right)^{x} \left(\frac{\mu}{\lambda+\mu}\right) \left[1 - e^{-(\lambda+\mu)\tau} \sum_{i=0}^{x} \frac{\left[(\lambda+\mu)\tau\right]^{i}}{i!}\right].$$
 (4)

These two scenarios also hold for the case of x = 0 but need to be augmented by an additional reason as to why no purchases could have occurred in the interval $(t, t + \tau]$: the individual was dead at the beginning of the interval, which occurs with probability

$$1 - e^{-\mu t}$$
. (5)

Combining (3)–(5) gives us the following expression for the probability of observing x purchases in the interval $(t, t + \tau]$, conditional on λ and μ :

$$P(X(t,t+\tau) = x \mid \lambda,\mu) = \delta_{x=0} \left[1 - e^{-\mu t} \right] + \frac{(\lambda \tau)^{x} e^{-\lambda \tau} e^{-\mu(t+\tau)}}{x!} + \left(\frac{\lambda}{\lambda+\mu}\right)^{x} \left(\frac{\mu}{\lambda+\mu}\right) e^{-\mu t} - \left(\frac{\lambda}{\lambda+\mu}\right)^{x} \left(\frac{\mu}{\lambda+\mu}\right) e^{-\lambda \tau} e^{-\mu(t+\tau)} \sum_{i=0}^{x} \frac{\left[(\lambda+\mu)\tau\right]^{i}}{i!}.$$
 (6)

4 Removing the Conditioning on λ and μ

In reality, we never know an individual's latent characteristics; we therefore remove the conditioning on λ and μ by taking the expectation of (6) over the distributions of Λ and M:

$$P(X(t,t+\tau) = x \mid r,\alpha,s,\beta) = \int_0^\infty \int_0^\infty P(X(t,t+\tau) = x \mid \lambda,\mu)g(\lambda \mid r,\alpha)g(\mu \mid s,\beta) \, d\lambda \, d\mu \,.$$
(7)

Substituting (1), (2), and (6) in (7) gives us

$$P(X(t,t+\tau) = x | r, \alpha, s, \beta) = \delta_{x=0} \mathsf{A}_1 + \mathsf{A}_2 + \mathsf{A}_3 - \sum_{i=0}^x \frac{\tau^i}{i!} \mathsf{A}_4$$
(8)

where

$$\mathsf{A}_{1} = \int_{0}^{\infty} \left[1 - e^{-\mu t} \right] g(\mu \mid s, \beta) \, d\mu \tag{9}$$

$$\mathsf{A}_2 = \int_0^\infty \int_0^\infty \frac{(\lambda \tau)^x e^{-\lambda \tau} e^{-\mu(t+\tau)}}{x!} g(\lambda \mid r, \alpha) g(\mu \mid s, \beta) \, d\lambda \, d\mu \tag{10}$$

$$\mathsf{A}_{3} = \int_{0}^{\infty} \int_{0}^{\infty} \left(\frac{\lambda}{\lambda+\mu}\right)^{x} \left(\frac{\mu}{\lambda+\mu}\right) e^{-\mu t} g(\lambda \mid r, \alpha) g(\mu \mid s, \beta) \, d\lambda \, d\mu \tag{11}$$

$$\mathsf{A}_{4} = \int_{0}^{\infty} \int_{0}^{\infty} \left(\frac{\lambda}{\lambda+\mu}\right)^{x} \left(\frac{\mu}{\lambda+\mu}\right) (\lambda+\mu)^{i} e^{-\lambda\tau} e^{-\mu(t+\tau)} g(\lambda \mid r, \alpha) g(\mu \mid s, \beta) \, d\lambda \, d\mu \tag{12}$$

Solving (9) and (10) is trivial:

$$\mathsf{A}_1 = 1 - \left(\frac{\beta}{\beta+t}\right)^s \tag{13}$$

$$\mathsf{A}_{2} = \frac{\Gamma(r+x)}{\Gamma(r)x!} \left(\frac{\alpha}{\alpha+\tau}\right)^{r} \left(\frac{\tau}{\alpha+\tau}\right)^{x} \left(\frac{\beta}{\beta+t+\tau}\right)^{s} \tag{14}$$

To solve (11), consider the transformation $Y = M/(\Lambda + M)$ and $Z = \Lambda + M$. Using the transformation technique (Casella and Berger 2002, Section 4.3, pp. 156–162; Mood et al. 1974, Section 6.2, p. 204ff), it follows that the joint distribution of Y and Z is

$$g(y, z \mid \alpha, \beta, r, s) = \frac{\alpha^r \beta^s}{\Gamma(r) \Gamma(s)} y^{s-1} (1-y)^{r-1} z^{r+s-1} e^{-z(\alpha - (\alpha - \beta)y)} .$$
(15)

Noting that the inverse of this transformation is $\lambda = (1 - y)z$ and $\mu = yz$, it follows that

$$\begin{aligned} \mathsf{A}_{3} &= \int_{0}^{1} \int_{0}^{\infty} y(1-y)^{x} e^{-yzt} g(y,z \mid \alpha,\beta,r,s) \, dz \, dy \\ &= \frac{\alpha^{r} \beta^{s}}{\Gamma(r)\Gamma(s)} \int_{0}^{1} \int_{0}^{\infty} y^{s} (1-y)^{r+x-1} z^{r+s-1} e^{-z(\alpha - (\alpha - (\beta + t))y)} \, dz \, dy \end{aligned}$$

$$= \frac{\alpha^{r}\beta^{s}}{\Gamma(r)\Gamma(s)} \int_{0}^{1} y^{s} (1-y)^{r+x-1} \left\{ \int_{0}^{\infty} z^{r+s-1} e^{-z(\alpha - (\alpha - (\beta + t))y)} dz \right\} dy$$

= $\alpha^{r}\beta^{s} \frac{\Gamma(r+s)}{\Gamma(r)\Gamma(s)} \int_{0}^{1} y^{s} (1-y)^{r+x-1} (\alpha - (\alpha - (\beta + t))y)^{-(r+s)} dy$
= $\frac{1}{B(r,s)} \frac{\alpha^{r}\beta^{s}}{\alpha^{r+s}} \int_{0}^{1} y^{s} (1-y)^{r+x-1} \left[1 - \left(\frac{\alpha - (\beta + t)}{\alpha} \right) y \right]^{-(r+s)} dy$

which, recalling Euler's integral for the Gaussian hypergeometric function,¹

$$= \frac{\alpha^r \beta^s}{\alpha^{r+s}} \frac{B(r+x,s+1)}{B(r,s)} {}_2F_1\left(r+s,s+1;r+s+x+1;\frac{\alpha-(\beta+t)}{\alpha}\right).$$
(16)

Looking closely at (16), we see that the argument of the Gaussian hypergeometric function, The boxing closely at (10), we see that the argument of the equation 1, p agreement 1, p agreement $2, \frac{\alpha - (\beta + t)}{\alpha}$, is guaranteed to be bounded between 0 and 1 when $\alpha \ge \beta + t$, thus ensuring convergence of the series representation of the function. However, when $\alpha < \beta + t$ we can be faced with the situation where $\frac{\alpha - (\beta + t)}{\alpha} < -1$, in which case the series is divergent. Applying the linear transformation (Abramowitz and Stegun 1972, equation 15.3.4)

$${}_{2}F_{1}(a,b;c;z) = (1-z)^{-a} {}_{2}F_{1}\left(a,c-b;c;\frac{z}{z-1}\right),$$
(17)

gives us

$$\mathsf{A}_{3} = \frac{\alpha^{r}\beta^{s}}{(\beta+t)^{r+s}} \frac{B(r+x,s+1)}{B(r,s)} {}_{2}F_{1}\left(r+s,r+x;r+s+x+1;\frac{\beta+t-\alpha}{\beta+t}\right).$$
(18)

We note that the argument of the above Gaussian hypergeometric function is bounded between 0 and 1 when $\alpha \leq \beta + t$. We therefore present (16) and (18) as solutions to (11), using (16) when $\alpha \geq \beta + t$ and (18) when $\alpha \leq \beta + t$. We can write this as

$$\mathsf{A}_3 = \alpha^r \beta^s \frac{B(r+x,s+1)}{B(r,s)} \mathsf{B}_1 \tag{19}$$

where

$$\mathsf{B}_{1} = \begin{cases} \frac{{}_{2}F_{1}\left(r+s,s+1;r+s+x+1;\frac{\alpha-(\beta+t)}{\alpha}\right)}{\alpha^{r+s}} & \text{if } \alpha \geq \beta+t \\ \frac{{}_{2}F_{1}\left(r+s,r+x;r+s+x+1;\frac{\beta+t-\alpha}{\beta+t}\right)}{(\beta+t)^{r+s}} & \text{if } \alpha \leq \beta+t \end{cases}$$
(20)

To solve (12), we also make use of the transformation $Y = M/(\Lambda + M)$ and $Z = \Lambda + M$. Given (15), it follows that

$$\begin{aligned} \mathsf{A}_{4} &= \int_{0}^{1} \int_{0}^{\infty} y(1-y)^{x} z^{i} e^{-(1-y)z\tau} e^{-yz(t+\tau)} g(y, z \mid \alpha, \beta, r, s) \, dz \, dy \\ &= \frac{\alpha^{r} \beta^{s}}{\Gamma(r)\Gamma(s)} \int_{0}^{1} \int_{0}^{\infty} y^{s} (1-y)^{r+x-1} z^{r+s+i-1} e^{-z(\alpha+\tau-(\alpha-(\beta+t))y)} \, dz \, dy \\ &= \frac{\alpha^{r} \beta^{s}}{\Gamma(r)\Gamma(s)} \int_{0}^{1} y^{s} (1-y)^{r+x-1} \left\{ \int_{0}^{\infty} z^{r+s+i-1} e^{-z(\alpha+\tau-(\alpha-(\beta+t))y)} \, dz \right\} \, dy \\ &= \alpha^{r} \beta^{s} \frac{\Gamma(r+s+i)}{\Gamma(r)\Gamma(s)} \int_{0}^{1} y^{s} (1-y)^{r+x-1} (\alpha+\tau-(\alpha-(\beta+t))y)^{-(r+s+i)} \, dy \\ &= \frac{\Gamma(r+s+i)}{\Gamma(r)\Gamma(s)} \frac{\alpha^{r} \beta^{s}}{(\alpha+\tau)^{r+s+i}} \int_{0}^{1} y^{s} (1-y)^{r+x-1} \left[1 - \left(\frac{\alpha-(\beta+t)}{\alpha+\tau} \right) y \right]^{-(r+s+i)} \, dy \end{aligned}$$

 ${}^{1}{}_{2}F_{1}(a,b;c;z) = \frac{1}{B(b,c-b)} \int_{0}^{1} t^{b-1} (1-t)^{c-b-1} (1-zt)^{-a} dt \, , \ c > b \, .$

which, recalling Euler's integral for the Gaussian hypergeometric function,

$$= \frac{\Gamma(r+s+i)}{\Gamma(r+s)} \frac{\alpha^r \beta^s}{(\alpha+\tau)^{r+s+i}} \frac{B(r+x,s+1)}{B(r,s)} \times {}_2F_1\left(r+s+i,s+1;r+s+x+1;\frac{\alpha-(\beta+t)}{\alpha+\tau}\right).$$
(21)

Noting that the argument of the Gaussian hypergeometric function is only guaranteed to be bounded between 0 and 1 when $\alpha \ge \beta + t \ (\forall \tau > 0)$, we apply the linear transformation (17), which gives us

$$A_4 = \frac{\Gamma(r+s+i)}{\Gamma(r+s)} \frac{\alpha^r \beta^s}{(\beta+t+\tau)^{r+s+i}} \frac{B(r+x,s+1)}{B(r,s)} \times {}_2F_1\left(r+s+i,r+x;r+s+x+1;\frac{\beta+t-\alpha}{\beta+t+\tau}\right),$$
(22)

The argument of the above Gaussian hypergeometric function is bounded between 0 and 1 when $\alpha \leq \beta + t \ (\forall \tau > 0)$. We therefore present (21) and (22) as solutions to (12): we use (21) when $\alpha \geq \beta + t$ and (22) when $\alpha \leq \beta + t$. We can write this as

$$\mathsf{A}_4 = \alpha^r \beta^s \frac{\Gamma(r+s+i)}{\Gamma(r+s)} \frac{B(r+x,s+1)}{B(r,s)} \mathsf{B}_2$$
(23)

where

$$\mathsf{B}_{2} = \begin{cases} \frac{{}_{2}F_{1}\left(r+s+i,s+1;r+s+x+1;\frac{\alpha-(\beta+t)}{\alpha+\tau}\right)}{(\alpha+\tau)^{r+s+i}} & \text{if } \alpha \geq \beta+t\\ \frac{{}_{2}F_{1}\left(r+s+i,r+x;r+s+x+1;\frac{\beta+t-\alpha}{\beta+t+\tau}\right)}{(\beta+t+\tau)^{r+s+i}} & \text{if } \alpha \leq \beta+t \end{cases}$$

$$(24)$$

Substituting (13), (14), (19), and (23) in (8) yields the following expression for the probability of observing x transactions in the time interval $(t, t + \tau]$:

$$P(X(t,t+\tau) = x | r, \alpha, s, \beta)$$

$$= \delta_{x=0} \left[1 - \left(\frac{\beta}{\beta+t}\right)^s \right] + \frac{\Gamma(r+x)}{\Gamma(r)x!} \left(\frac{\alpha}{\alpha+\tau}\right)^r \left(\frac{\tau}{\alpha+\tau}\right)^x \left(\frac{\beta}{\beta+t+\tau}\right)^s$$

$$+ \alpha^r \beta^s \frac{B(r+x,s+1)}{B(r,s)} \left\{ \mathsf{B}_1 - \sum_{i=0}^x \frac{\Gamma(r+s+i)}{\Gamma(r+s)} \frac{\tau^i}{i!} \mathsf{B}_2 \right\}$$
(25)

where expressions for B_1 and B_2 are given in (20) and (24), respectively.

We note that for t = 0, (25) reduces to the implied expression for $P(X(\tau) = x)$ as given in Fader and Hardie (2020, equation 16).

References

Abramowitz, Milton and Irene A. Stegun (eds.) (1972), *Handbook of Mathematical Functions*, New York: Dover Publications.

Casella, George, and Roger L. Berger (2002), *Statistical Inference*, 2nd edition, Pacific Grove, CA: Duxbury.

Fader, Peter S. and Bruce G.S. Hardie (2020), "Deriving an Expression for P(X(t) = x) Under the Pareto/NBD Model." ">http://brucehardie.com/notes/012/>

Mood, Alexander M., Franklin A. Graybill, and Duane C. Boes (1974), *Introduction to the Theory* of *Statistics*, 3rd edition, New York: McGraw-Hill Publishing Company.

Schmittlein, David C., Donald G. Morrison, and Richard Colombo (1987), "Counting Your Customers: Who Are They and What Will They Do Next?" *Management Science*, **33** (January), 1–24.