A note on a new class of generalized Pearson distribution arising from Michaelis-Menten function of enzyme kinetics

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Abstract

Many problems of enzyme kinetics can be described by a function known as the Michaelis-Menten (M-M) function. In this paper, motivated by the importance of Michaelis-Menten function in biochemical and other biological phenomena, we have introduced a new class of generalized Pearson distribution arising from Michaelis-Menten function. Various properties of this distribution are derived, for example, its probability density function (pdf), cumulative distribution function (cdf), moment, entropy function, and relationships with some well-known continuous probability distributions. The graphs of the pdf and cdf of our new distribution are provided for some selected values of the parameters. It is observed that our new distribution is positively skewed and unimodal. We hope that the findings of this paper will be useful in many applied research problems.

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1. Introduction

The Michaelis-Menten function is one of the most important mathematical functions to model many problems of biochemistry and other biological phenomena to describe enzymatic reactions, see, for example, Michaelis and Menten [13], Briggs and Haldane [3], Cleland [5], Fontes et al. [8], [9], among others. It is defined by the following equation:

\[ v = f ([S]) = \frac{V_{\text{max}} [S]}{K_m + [S]}, \]

where \( v \) is the initial velocity in an enzyme-catalyzed reaction, \( V_{\text{max}} \) is the maximal velocity, i.e. the velocity attended at very high concentration of substrate \([S]\), \( K_m \) is the Michaelis constant and corresponds to the concentration of substrate at which \( v = V_{\text{max}} / 2 \).

As pointed out by, Fontes et al. [9], the Michaelis-Menten equation (1) can be reduced to the following forms known as Type I, Type II and Type III respectively:

Type 1: \( y = \frac{ax}{b + x} \),

Type 2: \( y = \frac{n_0 + n_1 x}{d_0 + d_1 x} \),

Type 3: \( y = h + \frac{ax}{b + x} \),
where the constants have their usual meanings. For detailed mathematical analysis and applications to enzymatic reactions of the above three types of the Michaelis-Menten equation, see Fontes et al. [9], among others.

In this paper, motivated by the importance of Michaelis-Menten function in biochemistry and other biological phenomena, we have introduced a new class of generalized Pearson distribution arising from Michaelis-Menten function. The organization of this paper is as follows. Section 2 contains a review of existing classes of generalized Pearson continuous probability distributions as considered by various researchers. We have identified as many as 14 such distributions. Section 3 contains the derivations of the probability density function (pdf) and cumulative distribution function (cdf) of our proposed new class of generalized Pearson distribution, along with the graphs of the pdf and cdf for some selected values of the parameters, are provided. Section 5 contains some concluding remarks.

2. Review on existing classes of generalized Pearson system of distributions

A continuous probability distribution belongs to the Pearson system if, for a positive continuous random variable $X$, its probability density function (pdf) $f$ satisfies a differential equation of the form

$$\frac{1}{f(x)} \frac{df(x)}{dx} = \frac{x + a}{b x^2 + c x + d},$$

where $a$, $b$, $c$, and $d$ are real parameters such that $f$ is a pdf. The shapes of the pdf depend on the values of these parameters, based on which Pearson [16, 17] classified these distributions into a number of types known as Pearson Types I – VI. Later, in another paper, Pearson [18] defined more special cases and subtypes known as Pearson Types VII - XII. Many well-known distributions belong to these types of Pearson distributions which include Normal and Student’s $t$ distributions (Pearson Type VII), Beta distribution (Pearson Type I), Gamma distribution (Pearson Type III), among others. For details on the Pearson systems of continuous probability distributions, the interested readers are referred to Johnson et. al. [12]. In recent years, many researchers have considered a generalization of the Pearson system, known as generalized Pearson system of differential equation (GPE), given by

$$\frac{1}{f(x)} \frac{df(x)}{dx} = \sum_{j=0}^{m} a_j x^j \sum_{j=0}^{n} b_j x^j,$$

where $m$, $n \in \mathbb{N} \setminus \{0\}$ and the coefficients $a_j$ and $b_j$ are real parameters. The system of continuous univariate pdf’s generated by GPE is called a generalized Pearson system which includes a vast majority of continuous pdf’s by proper choices of these parameters. We have identified as many as 14 such distributions, which are provided below:

i) Roy [21] studied GPE, when $m = 2, n = 3, b_0 = 0$, to derive five frequency curves whose parameters depend on the first seven population moments.

ii) Dunning and Hanson [7] used GPE in his paper on generalized Pearson distributions and nonlinear programming.

iii) Cobb et al. [6] extended Pearson’s class of distributions to generate multimodal distributions by taking the polynomial in the numerator of GPE of degree higher than one and the denominator, say, $v(x)$, having one of the following forms:

a) $v(x) = 1, \ -\infty < x < \infty$,

b) $v(x) = x, \ 0 < x < \infty$,

c) $v(x) = x^2, \ 0 < x < \infty$,

d) $v(x) = x(1-x), \ 0 < x < 1$.

iv) Chaudhry and Ahmad [4] studied another class of generalized Pearson distributions when $m = 4, n = 3, b_0 = b_1 = b_2 = 0, \ a_4, a_3 = 2b_3 = 2\beta, b_3 \neq 0$.

v) Lefevre et al. [13] studied characterization problems based on some generalized Pearson distributions.

vi) Considering the following class of GPE

$$\frac{1}{f(x)} \frac{df_x(x)}{dx} = \frac{a_0 + a_1 x + a_2 x^2}{b_0 + b_1 x + b_2 x^2},$$

Sankaran [22] proposed a new class of probability distributions and established some characterization results based on a relationship between the failure rate and the conditional moments.

vii) Stavroyiannis and Stavroulakis [28] studied generalized Pearson distributions in the context of the superstatistics with non-linear forces and various distributions.
Rossani and Scarfone [20] have studied GPE in the following form
\[ \frac{1}{f(x)} \frac{df_x(x)}{dx} = -\frac{a_0 + a_1 x + a_2 x^2}{b_0 + b_1 x + b_2 x^2}, \]
and used it to generate generalized Pearson distributions in order to study charged particles interacting with an electric and/or a magnetic field.

Shakil et al. [24] defined a new class of generalized Pearson distributions based on the following differential equation
\[ \frac{df_x(x)}{dx} = \left(\frac{a_0 + a_1 x + a_2 x^2}{b_1 x}\right)f_x(x), \quad b_1 \neq 0, \]
which is a special case of the GPE (2), when \( m = 2, n = 1, \) and \( b_0 = 0 \). The solution to the differential equation (3) is given by
\[ f_x(x) = C x^{\alpha} \exp\left(-\mu x^2 - \beta x\right), \quad \alpha > 0, \beta > 0, \mu > 0, x > 0, \]
where \( \mu = -\frac{a_1}{2b_1}, \alpha = \frac{a_0}{b_1}, \beta = -\frac{a_2}{b_1}, b_1 \neq 0, \) and \( C \) is the normalizing constant given by
\[ C = \frac{\left(2^\mu\right)^{(\alpha+1)/2}}{\Gamma(\alpha+1) \exp\left(\beta^2/(8\mu)\right) D_{\alpha+1}(\beta/\sqrt{2\mu})}, \]
where \( D_p(z) \) denotes the well-known parabolic cylinder function.

Shakil and Kibria [23] consider the GPE (2) in the following form
\[ \frac{df_x(x)}{dx} = \left(\frac{a_0 + a_1 x + a_2 x^2}{b_1 x + b_{p+1} x^{p+1}}\right)f_x(x), \quad b_1 \neq 0, b_{p+1} \neq 0, x > 0, \]
when \( m = p, n = p + 1, a_1 = a_2 = \ldots = a_{p-1} = 0, \) and \( b_0 = b_1 = \ldots = b_p = 0 \). The solution to the differential equation (6) is given by
\[ f_x(x) = C x^{\alpha-1} \left(\alpha + \beta x^p\right)^{\nu}, \quad x > 0, \alpha > 0, \beta > 0, \mu > 0, \nu > 0, \text{ and } p > 0, \]
where \( \alpha = b_1, \beta = b_{p+1}, \mu = \frac{a_0 + b_1}{b_1}, \nu = \frac{a_1 b_{p+1} - a_2 b_1}{b_1 b_{p+1}}, b_1 \neq 0, b_{p+1} \neq 0, \) and \( C \) is the normalizing constant given by
\[ C = \frac{p^\mu (\alpha)^{\frac{n}{p}} (\beta)^{\frac{\nu}{p}}}{B\left(\frac{\mu}{p}, \nu - \frac{\mu}{p}\right)}, \]
where \( B\left(\frac{\mu}{p}, \nu - \frac{\mu}{p}\right) \) denotes the well-known beta function, and \( \nu > \frac{\mu}{p} \).

Shakil et al. [25] consider the GPE (2) in the following form
\[ \frac{df_x(x)}{dx} = \left(\frac{a_0 + a_1 x + a_2 x^2}{b_{p+1} x^{p+1}}\right)f_x(x), \quad b_{p+1} \neq 0, x > 0, \]
where \( m = 2p, n = p + 1, a_1 = a_2 = \ldots = a_{p-1} = a_{p+1} = \ldots = a_{2p-1} = 0, \) and \( b_0 = b_1 = b_2 = \ldots = b_p = 0 \). The solution to the differential equation (9) is given by
\[ f_x(x) = C x^{\alpha-1} \exp\left(-\alpha x^p - \beta x^p\right), \quad x > 0, \alpha \geq 0, \beta \geq 0, -\infty < \nu < \infty, \]
where \( \alpha = \frac{a_{2p}}{p b_{p+1}}, \beta = \frac{a_0}{p b_{p+1}}, \nu = \frac{a_1 + b_{p+1}}{b_{p+1}}, b_{p+1} \neq 0, p > 0, \) and \( C \) is the normalizing constant given by
\[ C = \frac{p^\mu (\alpha)^{\frac{n}{p}}}{K_{\frac{\mu}{p}}\left(2\sqrt{\alpha \beta}\right)}, \]
where \( K_{\frac{\mu}{p}}\left(2\sqrt{\alpha \beta}\right) \) denotes the well-known modified Bessel function of third kind. For the characterizations of the above continuous probability distribution, due to Shakil et al. [25], known in the literature as the Shakil-Kibria-Singh (SKS) distribution, the interested readers are referred to Hamedani [11] where the the Shakil-Kibria-Singh (SKS) distribution has been characterized by Hamedani [11] based on a simple relationship between two truncated moments, and the hazard function.

Hamedani [11] has defined a new variation of the continuous probability distribution (10) in a bounded domain. The pdf of Hamedani’s distribution is given by
where $\alpha > 0$, $\beta > 0$, and $p > 0$ are parameters and $C = \exp\left\{2\sqrt{\alpha\beta}\right\}$ is the normalizing constant.

The cdf corresponding to the pdf (12) is given by

$$F(x) = C \exp\left(-\alpha x^p - \beta x^{-p}\right), \quad 0 < x < \left(\frac{\beta}{\alpha}\right)^{1/p}.$$

For the special case of $\alpha = \beta$, we have

$$f(x) = \alpha p \exp\left(-p^{2} x^{-4p}\right)\left(1 - x^{-2p}\right)\exp\left(-\alpha(x^p + x^{-p})\right), \quad 0 < x < \left(\frac{\beta}{\alpha}\right)^{1/p},$$

where $\alpha > 0$ and $p > 0$ are parameters. As pointed out by Hamedani [11], the pdf $f$ given by (12) satisfies the following differential equation

$$\frac{1}{f(x)} \frac{df(x)}{dx} = \frac{\beta^2 p - \beta(p + 1)x^p - 2\alpha \beta p x^{2p} - \alpha(p - 1)x^{3p} + \alpha^2 p x^{4p}}{\beta x^{p+1} - \alpha x^{3p+1}},$$

which is a special case of GPE (2). For the characterization of the pdf in Eq. (12), when $p \in \mathbb{N} / \{0\}$, the interested readers are referred to Hamedani [11].

xiii) Ahsanullah et. al. [2] defined a new class of distributions as solutions of the GPE (2). They considered the following differential equation

$$\frac{df(x)}{dx} = \left(\frac{a_1 + a_2 x + a_3 x^2}{b_1 x^2 + b_2 x^3}\right)f(x),$$

which is a special case of the generalized Pearson Eq. (2) when $m = 2, n = 3$. Putting $b_3 = 1, b_4 = \gamma, a_1 = \beta \gamma, a_2 = \beta - \gamma + \gamma \nu, a_3 = \nu + \mu - 2, x > 0$, in (3), we have

$$\frac{1}{f(x)} \frac{df(x)}{dx} = \frac{\beta \gamma + (\beta - \gamma + \gamma \nu)x + (\nu + \mu - 2)x^2}{x^4 + \gamma x^3} = \frac{\nu - 1}{x} + \frac{\mu - 1}{x} + \frac{\beta}{x^4},$$

where we assume that $\beta > 0, \gamma > 0, 0 < \nu < 1, 0 < \mu < 1, -1 < \mu > 0$. Integrating the above equation, we have

$$f(x) = C x^{-1} (x + \gamma)^(\mu - 1) \exp\left(-\beta x^{-1}\right), \quad 0 < x < \infty,$$

Using the equation (3.471.7), Page 340 of Gradshteyn and Ryzhik [10], we easily obtain the following normalizing constant as

$$\frac{1}{C} = \beta^{\nu - \frac{1}{2}} \Gamma\left(1 - \mu - \nu\right) \exp\left(\frac{\beta}{2 \gamma}\right) W_{\nu, \mu - \nu - \frac{1}{2}}\left(\frac{\beta}{2 \gamma}\right),$$

where $W_{\nu, \mu} (z)$ denotes the well-known Whittaker function (see Abramowitz and Stegun [1], page 505, chapter 13).

xiv) Recently, Stavroyiannis [27] defined a new class of distributions as solutions of the GPE (2) by considering the following differential equation

$$\frac{1}{f(x)} \frac{df(x)}{dx} = \sum_{j=0}^{n} \frac{a_j x^j}{b_j x^j},$$

which is a special case of the generalized Pearson Eq. (2) when $m = 5, n = 6$. By taking special values of the coefficients $a_j$ and $b_j$, Stavroyiannis [27] obtained the GPE in the following form

$$\frac{1}{f(x)} \frac{df(x)}{dx} = \frac{a^2 \nu + 2a^2 m (x - \lambda) + a^2 \nu (x - \lambda)^2 + 2a^2 (1 + 2b) m (x - \lambda)^3 + ab \nu (x - \lambda)^4 + 4bm (x - \lambda)^5}{a^4 + (x - \lambda)^2 [a^2 + (x - \lambda)^2 + b (x - \lambda)^4]}$$

with its solution given by the following probability density function:

$$f(x) = C \exp\left[-\tan^{-1}\left(\frac{x - \lambda}{a}\right)\right],$$

where $\lambda$ is the location parameter, $a > 0$ is the scale parameter, $m > \frac{1}{2}$ and $b \geq 0$ control the kurtosis, $\nu$ is the asymmetry parameter, and $C$ is the normalizing constant. As pointed by Stavroyiannis [27], the above distribution
with the pdf (19) includes an extra fourth order term in the denominator to account for fat and thick-tails for the case of $b > 0$. The distribution becomes double peaked for the case of a negative $b$ coefficient, while for $b = 0$ the Pearson-IV distribution is regained. For details on these, the interested readers are referred to Stavroyiannis [27].

3. A new class of generalized pearson distribution arising from michaelis-menten function

In this section, for a positive continuous random variable $X$, we define a new class of generalized Pearson distributions based on the following differential equation

$$\frac{1}{f(x)} \frac{df(x)}{dx} = \frac{a_1 + b_1 x}{a_2 + b_2 x},$$

Equation (20)

$$f(x) = C \left(1 + \alpha x\right) \gamma \left(\alpha + \delta x\right)^{\theta} e^{-\gamma x}, \quad \alpha, \gamma, \delta > 0, \mu, \theta, \lambda \geq 0, x > 0,$$

Equation (21)

where $\mu = \frac{a_1 b_2}{b_1 b_2}$, $\alpha = a_2$, $\gamma = a_1$, $\delta = b_2$, $\theta = \frac{a_1}{b_2}$, $\lambda = \frac{b_1}{b_2}$, $(a_2, b_2 \neq 0)$, and $C$ denotes the normalizing constant.

3.1. Expressions for the normalizing constant

In order that the right side of the Eq. (21) represents a probability density function (pdf), we must have

$$\int_0^{\infty } f_x(x) dx = \int_0^{\infty } C \left(1 + \alpha x\right) \gamma \left(\alpha + \delta x\right)^{\theta} e^{-\gamma x} dx = 1.$$

Equation (22)

i) In Eq. (22), using twice the binomial series representation $(1+w)^\xi = \sum_{k=0}^{\infty } \left(\frac{s}{k!}\right) \left(-w\right)^{k}$, for any real value of $s$,

and Eq. 3.351.3/ P. 310 of Gradshteyn and Ryzhik [10], where $\left(\frac{s}{k}\right) = \frac{\Gamma(s + k)}{\Gamma(s)} = s(s + 1)...(s + k - 1)\left(s \neq 0\right)$,

and $\left(\frac{s}{k}\right) = 1$, denote the Pochhammer symbol, the normalizing constant $C$ is easily given by

$$C = \left[\sum_{k=0}^{\infty } \sum_{j=0}^{k} \left(-1\right)^{k-j} \left(\frac{\mu}{\gamma}\right)^{k+j} \frac{\gamma^{\xi+k+j} \lambda^{-k-j} (k+j)!}{k! j!} \right]^{-1}.$$

Equation (23)

ii) Again, in Eq. (22), using the binomial series representation $(1+w)^\xi = \sum_{k=0}^{\infty } \left(\frac{s}{k!}\right) \left(-w\right)^{k}$, for any real value of $s$,

and Equation 2.3.6.9 of Prudnikov et al., Vol. 1 [19], the expression for the normalizing constant $C$ is easily obtained, after simplification, as follows

$$C = \left[\sum_{j=0}^{\infty } \frac{\delta^{2-j} \alpha' \left(\frac{\gamma}{\delta}\right)^{j+1} \Psi\left(\left. j + 1, j + 2 - \theta; \frac{\lambda \gamma}{\delta}\right.\right)}{\Gamma(\frac{\gamma}{\delta})} \right]^{-1},$$

Equation (24)

where $\Psi(p, q; z) = \frac{1}{\Gamma(p)} \int_0^z t^{p-1} e^{-t} (1 + p)^{q-p-1} dt$ is known as Kummer’s (or degenerate hypergeometric) function of the second kind, see, for example, Abramowitz and Stegun [1], Gradshteyn and Ryzhik [10], and Oldham et. al. [15], among others.
3.2. Expression for the cumulative distribution function

Using twice the binomial series representation $(1 + w)^r = \sum_{k=0}^{\infty} \frac{r^k}{k!} (-w)^k$, for any real value of $s$, and Eq. 3.381.1/P. 317 of Gradshteyn and Ryzhik [10], the cumulative distribution function (cdf) of our new distribution is easily obtained as follows

$$F_X(x) = \int_0^x f_X(x) \, dx = C \sum_{k=0}^{\infty} \sum_{j=0}^{\infty} \frac{(-1)^k (\theta)_{k_j} (\mu), \gamma(\delta + j)}{\alpha! \lambda^{-k-j-1}} x^{k_j} \psi(k + j + 1, \lambda x)$$

(25)

where $\gamma(s, z) = \int_0^z t^{s-1} e^{-t} \, dt$ denotes the incomplete gamma function, and $C$ denote the normalizing constant given by the equation (23).

4. Distributional properties

In what follows, some properties of our proposed distribution are given below.

4.1. Graphs of the PDF and CDF

The possible shapes of the pdf $f(x)$ in Eq. (21) and the cdf $F(x)$ in Eq. (23) are provided for some selected values of the parameters in the following Figures 1-6. The effects of parameters can be easily seen from these graphs. Also, it is clear from these graphs that our proposed distributions of the random variable $X$ are positively (that is, right) skewed and unimodal.

![Fig. 1: PDF and Figure 2: CDF For $\mu = 1, 2, 3, 4$ when $\alpha = 1, \gamma = 1, \delta = 1, \theta = 0.5, \lambda = 0.5$.](image-url)
4.2. \textit{nth} Moment

In what follows, we derive the moments of our proposed distribution. We have

\[ E \left( X^n \right) = \int_0^\infty x^n f_X(x) \, dx = C \int_0^\infty x^n (1 + \alpha x) \mu (y + \delta x)^{-\theta} e^{-\lambda x} \, dx \quad (26) \]

In Eq. (26), using the binomial series representation \((1 + w)^{-r} = \sum_{k=0}^\infty \left( s \right)_k \frac{(-w)^k}{k!} \), for any real value of \( s \), and Eq. 2.3.6.9 of Prudnikov et al., Vol. 1 [19], the following expression for the \textit{nth} moment is easily obtained:

\[ E \left( X^n \right) = C \sum_{j=0}^{\infty} \frac{\delta^{-\theta} \mu^j \left( \frac{\delta}{\mu} \right)^{n+j+1-\theta}}{j!} \Gamma(n+j+1) \Psi \left( n+j+1, n+j+2-\theta; \frac{\lambda \delta}{\mu} \right) \quad (27) \]

where \( C \) denotes the normalizing constant given by (24). \( \Psi \left( p, q ; z \right) = \frac{1}{\Gamma(p)} \int_0^z t^{p-1} e^{-t} (1 + p)^{q-p-1} dt \) is known as Kummer’s (or degenerate hypergeometric) function of the second kind, and \( \Gamma(\cdot) \) denotes the gamma function defined by \( \Gamma(s) = \int_0^\infty t^{s-1} e^{-t} dt \), see, for example, Abramowitz and Stegun [1], Gradshteyn and Ryzhik [10], and Oldham et al.
[15], among others. Taking \( n = 1, 2, 3, \ldots \) in Eq. (27), we can easily obtain the moments of different orders, including the variance, \( \sigma^2 \), of our proposed distribution which can be obtained by using the formula: \( \sigma^2 = E(X^2) - (E(X))^2 \).

### 4.3. Shannon entropy

An entropy provides an excellent tool to quantify the amount of information (or uncertainty) contained in a random observation regarding its parent distribution (population). A large value of entropy implies the greater uncertainty in the data. As proposed by Shannon [26], if \( X \) is a none-negative continuous random variable with pdf \( f_x(x) \), then Shannon’s entropy of \( X \), denoted by \( h[f] \) or \( h(X) \), is defined as

\[
h[f] = h(X) = E[-\ln f(x)] = -\int_0^\infty f(x) \ln[f(x)] dx.
\]

Now, in Eq. (28) Above, using the pdf \( f(x) \) of the our proposed distribution, as given in Eq. (21), and then integrating and simplifying, Shannon entropy of our proposed distribution is easily obtained as follows

\[
h[f] = h(X) = -\ln C + \theta \ln \lambda - \lambda E(X) - \mu \sum_{i=1}^{\infty} \frac{(-1)^i}{i} E(X^i) + \theta \sum_{j=1}^{\infty} \frac{(-1)^j}{j} E(X^j),
\]

where \( C \) denotes the normalizing constant given by Eq. (24), and \( E(X) \), \( E(X^i) \) and \( E(X^j) \) denote the first, \( i \)th and \( j \)th moments respectively of our proposed distribution, and can be obtained by taking \( n = 1, i \) and \( j \) in Eq. (27) respectively.

### 4.4. Distributional relationships

It is easy to see that, by a simple transformation of the variable \( x \) or by taking special values of the parameters \( \{ \alpha, \gamma, \delta > 0; \mu, \theta, \lambda, \gamma > 0 \} \), number distributions are special cases of our proposed distribution as stated below.

i) Pearson III Distribution (when \( \theta = 0 \)): For the sake of motivation, the derivation of Pearson III distribution (when \( \theta = 0 \)) from our proposed distribution with PDF as given in (21) is provided below. Thus, when \( \theta = 0 \), we have

\[
x_i(x) = C (\alpha + \beta x)^{\gamma} e^{-\lambda x}, \quad \alpha > 0, \beta > 0, \mu > 0, x > 0, \lambda > 0,
\]

where \( \mu = -\frac{\alpha}{2b_1}, \alpha = a_1, \beta = -\frac{\alpha}{b_1}, b_1 \neq 0, \) and \( C \) is the normalizing constant given by

\[
C = \beta^{-\lambda} \lambda^{-m+1} e^{-\frac{\alpha}{\beta}} \left[ \Gamma(\mu + 1, \frac{\alpha \lambda}{\beta}) \right]^{-1},
\]

where \( \Gamma(s, z) \) denotes the incomplete gamma function.

The CDF of Pearson III distribution (when \( \theta = 0 \)) is given by

\[
F(x) = C \left( \frac{\beta}{\lambda} \right)^{\alpha} e^{\frac{\alpha}{\beta} x} \left[ \gamma(\mu + 1, \frac{\alpha \lambda}{\beta} + \lambda x) - \gamma(\mu + 1, \frac{\alpha \lambda}{\beta}) \right],
\]

where \( \gamma(s, z) \) denotes the incomplete gamma function, and \( C \) denote the normalizing constant as given above. The first, second, and \( n \)th moments of Pearson III distribution are obtained as follows:

First Moment: \( E(X) = C \beta^\alpha \sum_{m=0}^{\infty} \left( \frac{\mu}{m} \right) \left( \frac{\alpha}{\beta} \right)^m \frac{1}{\lambda^{m+1}} \Gamma(\mu - m + 2), \)

Second Moment: \( E(X^2) = C \beta^\alpha \sum_{m=0}^{\infty} \left( \frac{\mu}{m} \right) \left( \frac{\alpha}{\beta} \right)^m \frac{1}{\lambda^{m+2}} \Gamma(\mu - m + 3), \)

\( n \)th Moment: \( E(X^n) = C \beta^\alpha \sum_{m=0}^{\infty} \left( \frac{\mu}{m} \right) \left( \frac{\alpha}{\beta} \right)^m \frac{1}{\lambda^{m+n}} \Gamma(\mu - m + n), \)
where \( \Gamma(s) = \int_0^\infty t^{s-1}e^{-t}dt \) denotes the gamma function.

Entropy: \( h[f] = h(X) = -\ln C - \mu \ln \alpha + \lambda E(X) + \sum_{j=1}^{\infty} \frac{(-1)^j \mu^j}{\alpha^j} E(X^j) \),

where \( E(X) \) and \( E(X^j) \) denote the first and the \( j \)th moments of Pearson III distribution.

ii) Pearson VIII Distribution (when \( \mu = 0, \lambda = 0 \)).

iii) Pearson IX Distribution (when \( \lambda = 0, \theta = 0 \)).

iv) Pearson X Distribution (when \( \mu = 0, \theta = 0 \)).

v) A Special Case of Our Proposed Distribution (when \( \mu = 0 \)): Thus, when \( \mu = 0 \) in our proposed distribution with PDF as given in (21), we have

\[
f_x(x) = C (\gamma + \delta x)^{-\alpha} e^{-\lambda x}, \quad \alpha, \gamma, \delta > 0; \theta, \lambda \geq 0; x > 0,
\]

where \( C \) denotes the normalizing constant given by

\[
C = \left( \gamma \right)^{\alpha} \Psi\left( 1, 2 - \theta; \frac{2\gamma}{\delta} \right)^{-1},
\]

which is easily obtained by using Equation 2.3.6.9 of Prudnikov et al., Vol. 1 [19], where

\[
\Psi(p, q; z) = \frac{1}{\Gamma(p)} \int_0^\infty \frac{e^{-zt}t^{p-1}(1+t)^q^{-p-1} dt}{z}
\]

is known as Kummer’s (or degenerate hypergeometric) function of the second kind, see, for example, Abramowitz and Stegun [1], Gradshteyn and Ryzhik [10], and Oldham et. al. [15], among others.

vi) Distribution of the Product of the PDF’s of the Exponential and Some Members of the Family of Burr Distributions (Lomax, or Pareto Type I, or Pareto Type II): It is easy to see that, by a simple transformation of the variable \( x \) or by taking special values of the parameters \( \{\alpha, \gamma, \delta > 0; \theta, \lambda \geq 0\} \), the pdf of the above special case (v) of our proposed distribution (when \( \mu = 0 \)) can be expressed as the pdf of the product of the pdf’s of the exponential and some members of the family of Burr distributions (such as Lomax, or Pareto Type I, or Pareto Type II distributions).

5. Concluding remarks

In this paper, we have introduced a new class of generalized Pearson distribution arising from Michaelis-Menten function. Also, we have reviewed existing classes of continuous probability distributions which can be generated from the generalized Pearson system of differential equation (GPE), as given in Eq. (2). We have identified as many as fourteen such distributions. Various properties of our proposed distribution are derived, for example, its probability density function (pdf), cumulative distribution function (cdf), moment, entropy function, and relationships with some well-known continuous probability distributions. The graphs of the pdf and cdf of our proposed distribution are provided for some selected values of the parameters. It is observed that our proposed distribution is positively skewed and unimodal. We hope that the findings of this paper will be useful in many applied research problems. Some open problems and direction for future research for our proposed generalized Pearson distribution are characterization, estimation of parameters, applications to real world problems, Bayesian analysis, regression analysis, among others. Further, we hope that our proposed attempt will be helpful in designing a new approach of unifying different families of distributions based on the generalized Pearson differential equation. Some other open problems are following:

i) Can we unify all continuous probability distributions (known & unknown) through GPDE?

ii) Can we prove Existence & Uniqueness Theorem of Solutions for GPDE?

iii) Can we establish a Fixed Point Theorem for GPDE?

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Dedication
The authors would also like to dedicate this paper to Michaelis and Menten who published their seminal paper [Michaelis, L., and Menten, M. L. (1913). Die Kinetik der Invertinwirkung, Biochem. Z., 49, 333–369] on enzymatic reactions a century ago.

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