Continuous growth models in terms of generalized logarithm and exponential functions

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Abstract
Consider the one-parameter generalizations of the logarithmic and exponential functions which are obtained from the integration of non-symmetrical hyperboles. These generalizations coincide to the one obtained in the context of non-extensive thermostatistics. We show that these functions are suitable to describe and unify the great majority of continuous growth models, which we briefly review. Physical interpretation to the generalization function parameter is given for the Richards’ model, which has an underlying microscopic model to justify it.

Key words: population dynamics, generalized logarithmic and exponential functions, growth models, Richard’s model, Marusic and Bajzer’s model, Tsoularis and Wallace’s model

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1. Introduction
The convenience of generalizing the logarithmic and exponential functions has attracted the attention of researchers in the last years. One-parameter generalizations of logarithmic and exponential functions have been proposed in the several contexts [1,2,3,4,5,6,7]. More sophisticated generalizations have also been proposed such as a two and three-parameter [39], which includes all the others generalized logarithms as special cases. Examples of the convenience of these generalizations have been seen in different fields [10,11,12,13,14,15,16,17,18].

From an initial population size \( n_0 \), at an initial time \( t_0 \), growth models are used to predict the size \( n(t) \) of population at an arbitrary time \( t > t_0 \). The most important parameters are the intrinsic growth rate \( r \) and carrying capacity \( n_\infty = n(t \to \infty) \). Nevertheless frequently additional parameters are needed to adjust the model to a given experimental data. Up to date, a comprehensive description of continuous growth models is given by the Tsoularis and Wallace model [19], which generalizes the logistic-like models [20,21,22,23,24,25,26].

The main objective of this paper is to show that the one-parameter generalization of the logarithmic and exponential functions proposed [1] is suitable to describe continuous growth models. This paper is structured as follows. In Sec. 2 we present the one-parameter generalization of the logarithmic and exponential functions and a brief review of growth models. In Sec. 3 we consider a simple type of Bernoulli’s equation, which solution is the Zipf-Mandelbrot function. From the gradual generalization of the differential equations, their solutions retrieves the Richards’ growth model [22], Marusic and Bajzer model [26] and finally Tsoularis and Wallace model [19]. Since the Richards’ model has an underlying microscopic model [27], one is able to interpret physically the logarithmic function generalization parameter. Final remarks are presented in Sec. 4.
2. Brief review

To facilitate the connection between the one-parameter logarithmic and exponential generalizations and growth models, we briefly review these subjects in the following. We first present the simple geometrical arguments [28] which we have used to obtain a one-parameter generalization of the logarithmic and exponential functions. This natural geometrical generalization coincides exactly to the one proposed by Tsallis [1] in the context of non-extensive thermodynamics [29]. Instead of calling the extra parameter as \( q \), like in the precursor papers, we have used \( q = 1 - \tilde{q} \), which is more convenient to stress the symmetry properties of these functions. Next, we present some general continuous growth models and show how to retrieve the standard ones with the appropriate set of parameters.

2.1. The \( \tilde{q} \)-logarithmic, and the \( \tilde{q} \)-exponential functions

The \( \tilde{q} \)-logarithm function \( \ln_{\tilde{q}}(x) \) is defined as the value of the area underneath the non-symmetrical hyperbole \( f_q(t) = 1/t^{1-q} \) in the interval \( t \in [1, x] \), with \( x > 0 \):

\[
\ln_{\tilde{q}}(x) = \int_1^x \frac{dt}{t^{1-q}} = \lim_{\tilde{q} \to q} \frac{x^{\tilde{q}} - 1}{\tilde{q}} .
\]  

(1)

For any \( \tilde{q} \), the area is negative for \( 0 < x < 1 \), it vanishes for \( x = 1 [\ln_q(1) = 0] \), and it is positive for \( x > 1 \). This function is not the logarithmic function in the basis \( \tilde{q} \) [namely \( \log_{\tilde{q}}(x) \)] but a one-parameter generalization of the natural logarithmic function definition.

The \( \tilde{q} \)-exponential function \( e_{\tilde{q}}(x) \) is defined as the \( t \)-value, in such a way that the area underneath \( f_q(t) = 1/t^{1-q} \), in the interval \( t \in [1, e_{\tilde{q}}(x)] \), is \( x \). This is the inverse of the \( \tilde{q} \)-logarithmic function \( e_{\tilde{q}}[\ln_{q}(x)] = x = \ln_{\tilde{q}}[e_{\tilde{q}}(x)] \) and it is given by:

\[
e_{\tilde{q}}(x) = \lim_{\tilde{q} \to q} [1 + \tilde{q} x]^{1/\tilde{q}} .
\]  

(2)

Here the operator \( [a]_+ = \max(a, 0) \) is necessary since \( e_{\tilde{q}}(x) \) is not real if \( \tilde{q} x < 1 \). This is a non-negative function \( e_{\tilde{q}}(x) \geq 0 \) and \( x = 0 \) is a special point since \( e_{\tilde{q}}(0) = 1 \), independently of the \( \tilde{q} \) value.

2.2. Continuous growth models

Population dynamics deals with the evolution of species interacting with themselves and the environment. The one-species (logistic-like) models are well described in terms of growth models. Essentially, growth models connect the relative growth rate \( \alpha n(t)/|n(t)dt| = d\ln n(t)/dt \) of a population, with \( n(t) \) individuals of a given species at time \( t \), to an induced saturation function \( G_{r,n_\infty}(n) \): \( \ln n/\ln n = G_{r,n_\infty}(n) \), where \( n_\infty \) is the carrying capacity (environment saturation level), which corresponds to the asymptotic value of the stationary solution \( [\ln n(t)/dt = G_{r,n_\infty}(n) = 0] \) and \( r \) is the intrinsic growth rate (individual contribution). To compare populations in different environments, a better variable to work with is the ratio between the population size and the carrying capacity (final population size): \( p(t) = n(t)/n_\infty \). In this new variable, one has:

\[
\frac{d\ln p}{dt} = G_r(p) ,
\]  

(3)

where we notice that now, the induced saturation function depends only on \( r \), which we write \( G_r \propto r \). Thus, the environmental contribution is taken into account into the variable itself.

In the following we consider one-species (logistic-like) models. The Malthus’ (exponential) model is the simplest one \( [G_r(p) = r] \), which is extended basically to two other models:

- the Verhulst (logistic) model \( [G_r(p) = r(1-p)] \), whose discretization leads to the logistic map (prototype example in dynamical system) and
- the Gompertz’s model \( [G_r(p) = -\kappa \ln p] \), which is also used to calculate life insurances taxes [20,21].

It is interesting to observe that in Ref. [30, p. 83], it is shown that the Verhulst and Gompertz models can be generalized by a simple function proposed by Montroll and Badger [31], which in fact is exactly the Richards’ model [22].

\[
G_\kappa(p) = \kappa \frac{1 - p^{\tilde{q}}(t)}{\tilde{q}} ,
\]  

(4)

where \( \kappa = r \tilde{q} \). When \( \tilde{q} = 1 \), one obtains the Verhulst model and when \( \tilde{q} \to 0 \), one retrieves the Gompertz’s model. It is interesting to point out that \( \kappa \) and \( r \) are not exactly the same intrinsic growth parameter, although they are linearly related.

A more general description of growth models has been given recently by the Tsoularis and Wallace model [19, p. 22]. They considered the following induced saturation function:

\[
G_\kappa(p) = \kappa p^{a-1}(t) \left[ \frac{1 - p^{\tilde{q}}(t)}{\tilde{q}} \right] ^\gamma ,
\]  

(5)

where \( \kappa = r \tilde{q} \gamma n_\infty ^{\alpha -1} \). The formal solution of Eq. (3) with the above induced saturation function is given by:

\[
B_{p^a(t)} \left( \frac{1 - \alpha}{\tilde{q}}, 1 - \gamma \right) = B_{p^\infty} \left( \frac{1 - \alpha}{\tilde{q}}, 1 - \gamma \right) = nt ,
\]  

(6)

where \( B_{p^a(t)} = \int_0^t dt \frac{\alpha - 1}{\alpha - 1} (1-t)^{-b} \) is the incomplete beta function and two sets of values are acceptable:

(i) \( \alpha < 1, \tilde{q} > 0 \) and \( \gamma < 1 \),
(ii) \( \alpha > 1, \tilde{q} < 0 \) and \( \gamma < 1 \).

Consider the special cases of the above model:

- For \( \gamma = 0 \), one has Zipf-Mandelbrot-like growth models, which has two special cases:
  - exponential, for \( \alpha = 1 \) and
  - power-law, for \( \alpha > 1 \).
The derivative of the \( \tilde{q} \)-exponential function with respect to \( x \) is: \( d\tilde{q}/(k x) = k \ [\tilde{q}(k x)]^{1-\tilde{q}} \), so that the \( \tilde{q} \)-exponential function is the solution of the following nonlinear first order differential equation: \[ \frac{dy(x)}{dx} = k \ y^{1-\tilde{q}}(x) , \] where \( k \) is a constant. In Chemistry it is known as kinetic equation of order \( 1-\tilde{q} \). This equation is not directly related to any growth model, but if one thinks that \( \tilde{y} = \ln(p) \), then one has the hyper-Gompertz model. The solution of Eq. (7) can be found rewriting it as \[ \frac{dy(x)}{dx} = m y(x) + (k - m) y^{1-\tilde{q}}(x) , \] which leads to the Richard’s model as we will show.

3. \( \tilde{q} \)-generalized functions and growth models

In what follows we will first deal with a particular type of Bernoulli’s equation \( [d/dx + p(x)]y(x) = q(x) y^{1-\tilde{q}}(x) \), with \( p(x) = 0 \) and \( q(x) = 1 \). We show that this equation (kinetic equation of arbitrary order) is related to the hyper-Gompertz model. When a new linear term is added to this equation, it represents the Richard’s growth models and in this case we have been able to give a physical interpretation to \( \tilde{q} \). If this new added term is non-linear, we obtain an equation, which represents the Marusic-Bajzer models. The effort term can be included considering the \( \tilde{q} \)-logarithm function raised to a power, which gives rises to the Tsoularis and Wallace model.

3.1. Zipf-Mandelbrot model

The derivative of the \( \tilde{q} \)-exponential function with respect to \( x \) is: \( d\tilde{q}(k x)/dx = k \ [\tilde{q}(k x)]^{1-\tilde{q}} \), so that the \( \tilde{q} \)-exponential function is the solution of the following nonlinear first order differential equation: \[ \frac{dy(x)}{dx} = k \ y^{1-\tilde{q}}(x) , \] where \( k \) is a constant. In Chemistry it is known as kinetic equation of order \( 1-\tilde{q} \). This equation is not directly related to any growth model, but if one thinks that \( \tilde{y} = \ln(p) \), then one has the hyper-Gompertz model. The solution of Eq. (7) can be found rewriting it as \[ \frac{dy(x)}{dx} = m y(x) + (k - m) y^{1-\tilde{q}}(x) , \] which leads to the Richard’s model as we will show.

3.2. Richards-like growth models

Eq. (7) can be generalized to take care of other interesting situations. The following differential equation:

\[ \frac{dy(x)}{dx} = m y(x) + (k - m) y^{1-\tilde{q}}(x) , \] has been used in growth models [26] and other contexts [23]. Factoring \( m y(x) \) and calling \( \tilde{y} = [m/(m - k)]^{1/\tilde{q}} y(x) \), Eq. (10) is written as:

\[ \frac{d\ln \tilde{y}^{-1}(x)}{dx} = m \tilde{q} \ln \tilde{y}^{-1}(x) , \]

which leads to the Richard’s model as we will show.

3.2.1. Richards’ growth model

In Eq. (11), calling \( t \) the independent variable, \( p = 1/\tilde{y} \) and \( \tilde{m} = -\kappa \), one rewrites it as:

\[ \frac{d\ln p(t)}{dt} = -\kappa \ln p(t) , \]

which solution is

\[ p(t) = \frac{1}{e^{\tilde{q}[-\ln \tilde{q}(p_0^{-1})e^{-\kappa t}]}} \]

\[ = e^{-\tilde{q}[-\ln \tilde{q}(p_0^{-1})e^{-\kappa t}]}} . \]

This is the Richards’ growth model [22].

The discretization of Eq. (11) leads to a generalization of the logistic map different from the previously considered one (Feigenbaum’s map). An initial study of this recursive equation has been presented in Ref. [34].

3.2.1.1. Particular Cases. Some particular cases can of Eqs. (11) and (12) be retrieved:

- If \( \tilde{q} = 1 \), one recovers the Verhulst’s model (logistic equation):

\[ \frac{d\ln p}{dt} = -\kappa \ln \left(p\right) = \kappa (1 - p) , \]

which has the following solution:

\[ p(t) = e^{-1[(1 - p_0^{-1})e^{-\kappa t}] = \frac{1}{1 + (p_0^{-1} - 1)e^{-\kappa t}} , \]

where \( p_0 = n_0/n_\infty \) with \( n_0 = n(0) \) being the initial population size and \( n_\infty = n_\infty \) the carrying capacity.

- If \( \kappa > 0 \), then for \( t \gg 1/\kappa, \kappa(\infty) = 1 \), i.e. the carrying capacity is an attractor.

- If \( \kappa < 0 \), the carrying capacity \( n_\infty \) becomes a repulsor and the solutions \( p(t) = 1/[1 + (p_0^{-1} - 1)e^{\kappa t}] \), either converge, for \( n_0 < n_\infty \) (i.e. \( p_0 < 1 \)) or diverge, otherwise. This divergence means that the population becomes non-limited for times greater than \( t^* = \ln(1 - p_0^{-1})/\kappa \), which is solution of \( 1 + (p_0^{-1} - 1)e^{\kappa t^*} = 0 \) and depends on the initial value \( p_0 \).
As \( n_{\infty} \to \infty \), then \( p \to 0 \), and one has:

\[
\frac{d \ln p}{dt} = \kappa, \tag{15}
\]

which is the Malthus’ model. The solution of Eq. (15) is:

\[
p(t) = p_0 e^{\kappa t}. \tag{16}
\]

- If \( \tilde{q} = 0 \), one retrieves the Gompertz’s model:

\[
\frac{d \ln p}{dt} = -\kappa \ln p, \tag{17}
\]

which solution is:

\[
p(t) = p_0 e^{-\kappa t}. \tag{18}
\]

- If \( \tilde{q} = -1 \), one has the monomolecular or Mitscherlich growth model [19, p. 33]:

\[
\frac{dp(t)}{dt} = k[1 - p(t)], \tag{19}
\]

so that

\[
p(t) = 1 - (1 - p_0) e^{-kt}. \tag{20}
\]

Observe that Mitscherlich model [Eq. (19)], which uses \( \frac{dp}{dt} \) differs fundamentally from Verhulst model [Eq. (13)], which uses \( \frac{d \ln p}{dt} \).

3.2.1.2. Microscopic Model. The competition between cell drive to replicate and inhibitory interactions, that are modeled by long range interaction among the cell, furnishes an interesting microscopic mechanism to obtain Richards’ model [27]. The long range interaction is dependent on the distance \( r \) between two cells as a power law \( r^\gamma \) and the cells have a fractal structure characterized by a fractal dimension \( D_f \). Using Eq. (1) one can write their Eq. (7) of Ref. [27] as:

\[
\ln n(t)/dt = \langle G \rangle - J \omega \ln q[D_f n(t)/\omega],
\]

where \( \omega \) is a constant related to geometry of the problem and \( \tilde{q} = 1 - 1/D_f \). Here the parameter \( \tilde{q} \) acquires a physical meaning related to the interaction range \( \gamma \) and fractal dimension of the cellular structure \( D_f \). This physical interpretation of \( \tilde{q} \) has only been possible due to Richards’ model underlying microscopic description.

Another point of view is to obtain particular cases from the microscopic model of Ref. [27]. If all the cells interact with the same intensity for a fixed \( D_f \neq 0 \), then \( \gamma = 0 \) and \( \tilde{q} = 1 \), so that Eqs. (11) and (12) lead to the Verhulst model. Another way to obtain the Verhulst model is to fix the cell interaction range \( \gamma \neq 0 \) and consider large fractal dimensionality \( D_f \to \infty \). Both cases illustrates that the Verhulst model is a mean field model. If cell interaction range matches with the cell fractal structure then \( \gamma = D_f \) and \( \tilde{q} = 0 \), so that Eqs. (11) and (12) lead to the Gompertz’s model.

If \( \gamma > D_f \), then \( \tilde{q} < 0 \) and Eq. (12) reads:

\[
p(t) = 1 + \ln \tilde{q}(p_0) e^{-\kappa t}; \text{ for } \kappa t \gg 1, \text{ which leads to exponential behavior. In particular if } \kappa = 2D_f, \text{ one retrieves the Mitscherlich model [Eq. (20)].}
\]

3.2.2. Richard-Schaefer’s model

If individuals can be inserted to or removed from a given population at fixed time intervals at a rate \( \epsilon \), which is called the effort rate, then one has the Schaefer’s model [20,21]:

\[
\frac{d \ln p}{dt} = -\kappa \ln p - \epsilon. \tag{21}
\]

The solution of Eq. (21) is:

\[
p_{\epsilon,\kappa}(t) = \frac{e_{\tilde{q}}(\epsilon)}{e_{\tilde{q}}(\epsilon_0)} e^{-\kappa \epsilon t} \tag{22}
\]

so that when \( \epsilon = 0 \), one recovers Eq. (12).

3.3. Marusic and Bajzer’s growth model

As proposed by M. Marusic and Z. Bajzer [26] the equation

\[
\frac{dy(x)}{dx} = ay^a(x) + by^b(x), \tag{23}
\]

with \( a, b, \alpha, \beta \in \mathbb{R} \), furnishes the basis of the models of one-species dynamical population, which generalizes traditional growth models.

We remark that Eq. (23) with \( a > 0 \) and \( b < 0 \) was first studied in Ref. [35], which deals with qualitative laws in metabolism and growth. A more fundamental derivation of Eq. (23) has been furnished in Ref. [37,38] in the allometric morphogenesis (form or pattern generation) context. Marusic and Bajzer have considered arbitrary real parameters and solved in Eq. (23).

Further, we notice that one can rewrite Eq. (23) as:

\[
dy(x)/dx = ay^a(x) + by^b(x)(\beta - \alpha)[y^{\beta-a}(x) - 1 + 1/\beta - \alpha] = \tilde{a}y^a(x) + \tilde{b}y^b(x), \tag{24}
\]

where \( \tilde{a} = a + b \) and \( \tilde{b} = b(\beta - \alpha) \). In this way (Eq. (2)) in the Marusic and Bajzer paper [26] is generalized and retrieved for the particular case \( \beta \to \alpha \).

Calling \( a = m, b = k - m, \alpha = 1 + \tilde{q}' \) and \( \beta = 1 + \tilde{q}' - \tilde{q} \), in Eq. (23), one obtains:

\[
\frac{dy(x)}{dx} = my^{1+\tilde{q}'}(x) + (k - m)y^{1+\tilde{q}'} - \tilde{q}(x). \tag{25}
\]

Taking \( \tilde{q}' = 0 \), one recovers Eq. (19). Factoring \( my^{1+\tilde{q}'}(x) \) and calling \( \tilde{y} = [m/(m - k)]^{1/q} y(x) \), Eq. (23) is written as:

\[
\frac{d \ln \tilde{y}(x)}{dx} = m\tilde{q} \left( \frac{k - m}{m} \right)^{\tilde{q}/\tilde{q}} \ln \frac{\tilde{y}(x)}{\tilde{y}(x)}. \tag{26}
\]

Calling \( t \) the independent variable \( x, p = 1/\tilde{y} \) and \( m\tilde{q}[k - m]/m)^{\tilde{q}/4} = \kappa \), one rewrites Eq. (23) as:

\[
\frac{d \ln \tilde{q}(p(t))}{dt} = -\kappa \ln \tilde{q}(p(t)). \tag{26}
\]

This equation has analytical solutions only for some particular values of \( \tilde{q}' \). For \( \tilde{q}' = 0 \), one recovers the Richards’ model [Eq. (11)]. For \( \tilde{q} \neq 0 \), other common growth models are obtained as limiting cases, as we see below.
3.3.1. Generalized von Bertalanffy’s model

For \( \tilde{q}' = \tilde{q} \), one has the generalized von Bertalanffy model [19 p. 47], and Eq. (26) has the following solution:

\[
p(t) = e_\tilde{q}[\ln q(p_0) e^{-\kappa t}]. \tag{27}
\]

If \( \tilde{q} = 1/3 \), then \( \tilde{q}' = \tilde{q} = 1/3 \) and one recovers the specialized von Bertalanffy model [36].

3.3.2. Smith’s model

The Smith’s model [19 p. 35] replaces the term \( 1 - p \) in Verhulst model by \( 1 - (p + a \, dp/dt) \). The resulting equation is

\[
\frac{d\ln p(t)}{dt} = r \left[ \frac{1 - p(t)}{1 + r a p(t)} \right], \tag{28}
\]

which cannot be obtained immediately from Eq. (26), but from an approximation of it by setting \( \tilde{q} = 1 \) and \( \tilde{q}' = 1 - 0.473 \).

3.4. Tsoularis and Wallace’s growth model

The growth model proposed by A. Tsoularis and J. Wallace [19] [given by Eq. 6] can be obtained generalizing Eq. (26) to:

\[
\frac{d\ln q(p(t)}{dt} = \kappa \left[ - \ln q(p(t) \right]^{\gamma}. \tag{29}
\]

which solution is given by Eq. (6), with \( \tilde{q} = \tilde{q}(\gamma) \). For \( \gamma = 1 \), one retrieves the Marusic and Bajzer growth model [Eq. (26)]. For \( \gamma \neq 1 \), other common growth models are retrieved as limiting cases.

The parameter \( \gamma \) has the effect of an effort rate. This can be seen rewriting Eq. (29) in terms of a supplementary \( \tilde{q} \)-logarithmic function:

\[
\frac{d\ln \tilde{q} p(t)}{dt} = \tilde{k} \ln_\tilde{q} \left[ - \ln q(p(t)] - \tilde{\varepsilon}, \tag{30}
\]

where \( \tilde{k} = \kappa \gamma \) and \( \tilde{\varepsilon} = -\kappa \). For comparison, see Eq. (21).

3.4.1. Hyper-Gompertz’s model (or Exponential polynomial)

For \( \tilde{q}' = \tilde{q} = 0 \) in Eq. (29), one finds the hyper-Gompertz’s model

\[
\frac{d\ln p(t)}{dt} = \kappa \left[ - \ln p \right]^{\gamma}. \tag{31}
\]

Calling \( y = \ln p \), this equation becomes Eq. (7), whose solution is given by Eq. (5). In the original variable, the solution of Eq. (31) is:

\[
p(t) = p_0^{e^{\gamma(-\kappa t/(\ln p_0)^{\gamma})}} = e^{\ln(p_0)e^{-\gamma(-\kappa t/(\ln p_0)^{\gamma})}}, \tag{32}
\]

where we point out the presence of a generalized exponential function in the argument of an ordinary one.

3.4.2. Blumberg’s model

For \( \tilde{q} = 1 \) in Eq. (29), one retrieves the Blumberg model [24], which has been proposed to model organ size evolution:

\[
\frac{d\ln \tilde{q} p}{dt} = \kappa \left[ - \ln_1(p) \right]^{\gamma} \\
\frac{dp}{dt} = \kappa p^{1-q'}(1-p)^\gamma, \tag{33}
\]

which can be rewritten as:

\[
p(t) = \int p^{\tilde{q} - 1}(1-p)^{-\gamma} = \kappa t. \tag{34}
\]

If \( \tilde{q} > 0 \) and \( \gamma < 1 \), one has a closed analytical solution [as in Eq. (6), \( \alpha = 1 - \tilde{q} \)]:

\[
B_{p(t)}(\tilde{q}', 1 - \gamma) - B_{p_0}(\tilde{q}', 1 - \gamma) = \kappa t. \tag{35}
\]

Notice that in Eq. (33) for:

- \( \gamma = 0 \), one has the power law induced saturation function \( G_\kappa(p) = \kappa p^{1-\tilde{q}} \) and
- \( \gamma = 1 + 1/N \) and \( \tilde{q} = 1 - 1/N \), one has the hyperbolic model for regenerative growth (see Ref. [23]).

3.4.3. Turner et al. model (Generic growth)

For \( \tilde{q}' = \tilde{q}(\gamma - 1) \) in Eq. (29), one retrieves the Turner et al. model [25]:

\[
\frac{d\ln p(t)}{dt} = \kappa p^{0(1-\gamma)(\gamma) \left[ - \ln q(p(t) \right]^{\gamma}, \tag{36}
\]

which has the following analytical solution:

\[
p(t) = e^{-\tilde{q}(\ln q(p_0) e_1-\gamma \left[ -\kappa t (\ln q(p_0)^{-1})^{-1} \right])}. \tag{37}
\]

The Turner et al. model can be solved analytically only because of the judicious choice of \( \tilde{q}' \).

3.4.4. Retrieving growth models

As one can see, all continuous growth models considered in this paper are generalized by Eq. (29). To recover those models, take appropriate values of \( \tilde{q}' \), \( \tilde{q} \), \( \gamma \) and \( \kappa \) shown in Tab. 1. Some particular cases have analytical solution and they are shown in Fig. 1. It can be observed that for the Richard-Schaefer model, the asymptotic value is \( p_\infty = e_\tilde{q}(\epsilon) = \sqrt{0.8} = 0.89 . . . \), due to the effort \( \epsilon = -0.1 \) rate, instead of \( p_\infty = 1 \).

4. Conclusion

The main objective of this paper is to show that the one-parameter generalization of the logarithmic and exponential function obtained from non-extensive statistical physics is suitable to describe a large class of continuous population dynamics (growth) model. On one hand, the \( \tilde{q} \)-logarithmic function appears in the differential equations that define
the models. On the other hand, the $\tilde{q}$-exponential function appears in the solutions of those differential equations, when it is analytically available. In the presentation we have gradually increased the number of parameters of the differential equation to establish connection with growth models. Eq. (29) represents the most general model, the one of Tsoularis and Wallace. It has a simple form when written in terms of $\tilde{q}$-logarithmic function. A physical interpretation has been given to $\tilde{q}$ in the Richards’ model. This has been possible due to the microscopic model presented in Ref. [27].

| Model                        | $\tilde{q}'$ | $\tilde{q}$ | $\gamma$ | $\kappa$ | Equation               |
|------------------------------|--------------|-------------|----------|----------|------------------------|
| Malthus (exponential)        | 0            | *           | 0        | r        | $\frac{dp}{dt} = r$   |
| Verhulst (logistic)          | 0            | 1           | 1        | $\kappa\ln p$ | $\frac{dp}{dt} = r(1-p)$ |
| Gompertz                     | 0            | 1           | *        | $\kappa\ln p$ | $\frac{dp}{dt} = -\kappa\ln p$ |
| hyper-Gompertz               | 0            | 0           | *        | $\kappa\ln p$ | $\frac{dp}{dt} = \kappa(-\ln p)^\gamma$ |
| Richards                     | 0            | *           | 1        | $\kappa\ln p$ | $\frac{dp}{dt} = r(1-p^\kappa)$ |
| Tsoularis and Wallace        | *            | *           | *        | $\kappa\ln p$ | $\frac{dp}{dt} = \kappa(-\ln p)^\gamma$ |
| Marusic and Bajzer           | *            | *           | 1        | $\kappa\ln p$ | $\frac{dp}{dt} = -\kappa\ln p$ |
| Mitscherlich (monomolecular) | 1            | 1           | 1        | $\kappa\ln p$ | $\frac{dp}{dt} = \kappa(1-p)^\gamma$ |
| Blumberg                     | *            | 1           | 1        | $\kappa\ln p$ | $\frac{dp}{dt} = \kappa(1-p)^\gamma$ |
| Turner et al.                | $1 + \tilde{q}(1-\gamma)$ | 1           | 1        | $\kappa\ln p$ | $\frac{dp}{dt} = \kappa\tilde{q}^{(1-\gamma)}(-\ln p)^\gamma$ |
| Specialized von Bertalanffy  | $1/3$        | 1/3         | *        | $\kappa\ln p$ | $\frac{dp}{dt} = \kappa\ln_{1/3} p$ |
| Generalized von Bertalanffy  | $\tilde{q}$ | *           | 1        | $\kappa\ln p$ | $\frac{dp}{dt} = \kappa\ln_{1/3} p$ |
| Smith                        | $1 - 0.473$  | 1           | 1        | *        | approximation         |

Table 1
Growth models which can be obtained from Eq. (29). The asterisk means that the considered parameter can take an arbitrary value within its domain.

Fig. 1. Models which present analytical solution from Eq. (29). The curves have been obtained with the following values: $p_0 = 0.001$, $a = 0.1$ and $\epsilon = -0.1$. For parameters of arbitrary value in Tab. 1 we have taken: $\kappa = 1$, $q = 2$, and $\gamma = 1.5$.

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