# On Continuous Limiting Behaviour for the $q(n)$-Binomial Distribution with $q(n) \rightarrow 1$ as $n \rightarrow \infty$ 

Malvina Vamvakari<br>Department of Informatics and Telematics, Harokopio University of Athens, Athens, Greece<br>Email: mvamv@hua.gr

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#### Abstract

Recently, Kyriakoussis and Vamvakari [1] have established a $q$-analogue of the Stirling type for $q$-constant which have lead them to the proof of the pointwise convergence of the $q$-binomial distribution to a Stieltjes-Wigert continuous distribution. In the present article, assuming $q(n)$ a sequence of $n$ with $q(n) \rightarrow 1$ as $n \rightarrow \infty$, the study of the affect of this assumption to the $q(n)$-analogue of the Stirling type and to the asymptotic behaviour of the $q(n)$-Binomial distribution is presented. Specifically, a $q(n)$ analogue of the Stirling type is provided which leads to the proof of deformed Gaussian limiting behaviour for the $q(n)$-Binomial distribution. Further, figures using the program MAPLE are presented, indicating the accuracy of the established distribution convergence even for moderate values of $n$.


Keywords: Stirling Formula; $q(n)$-Factorial Number of Order $n$; Saddle Point Method; $q(n)$-Binomial Distribution; Pointwise Convergence; Gauss Distribution

## 1. Introduction and Preliminaries

In last years, many authors have studied $q$-analogues of the binomial distribution (see among others [2-4]). Specifically, Kemp and Kemp [3] defined a $q$-analogue of the binomial distribution with probability function in the form

$$
\begin{align*}
f_{X}(x) & =P(X=x) \\
& =\binom{n}{x}_{q} q^{\binom{x}{2}} \theta^{x} \prod_{j=1}^{n}\left(1+\theta q^{j-1}\right)^{-1},  \tag{1}\\
x & =0,1, \cdots, n,
\end{align*}
$$

where $\theta>0,0<q<1$, by replacing the loglinear relationship for the Bernoulli probabilities in Poissonian random sampling with loglinear odds relationship. Also, Kemp [4] defined (1) as a steady state distribution of birth-abort-death process.

Futhermore, Charalambides [2] considering a sequence of independent Bernoulli trials and assuming that the odds of success at the $i$ th trial given by

$$
\pi_{i}=\theta q^{i-1}, i=1,2, \cdots, 0<q<1,0<\theta<\infty,
$$

is a geometrically decreasing sequence with rate $q$, de-
rived that the probability function of the number $X$ of successes up to $n$-trail is the $q$-analogue of the binomial distribution with p.f. given by Equation (1).

For $q$ constant, the $q$-binomial distribution has finite mean and variance when $n \rightarrow \infty$. Thus, the asymptotic normality in the sense of the DeMoivre-Laplace classical limit theorem did not conclude, as in the case of ordinary hypergeometric series discrete distributions. Also, asymptotic methods-central or/and local limit theorems-are not applied as in Bender [5], Canfield [6], Flajolet and Soria [7], Odlyzko [8] et al.

Recently, Kyriakoussis and Vamvakari [1], for $q$ constant, established a limit theorem for the $q$-binomial distribution by a pointwise convergence in a $q$-analogue sense of the DeMoivre-Laplace classical limit theorem. Specifically, the pointwise convergence of the $q$-binomial distribution to a Stieltjes-Wigert continuous distribution was proved. In detail, transferred from the random variable $X$ of the $q$-binomial distribution (1) to the equal-distributed deformed random variable $Y=[X]_{1 / q}$, then, for $n \rightarrow \infty$, the $q$-binomial distribution was approximated by a deformed standardized continuous Stieltjes-Wigert distribution as follows

$$
\begin{align*}
& f_{X}(x) \cong \frac{q^{1 / 8}\left(\log q^{-1}\right)^{1 / 2}}{(2 \pi)^{1 / 2}} \\
& \cdot\left(q^{-3 / 2}(1-q)^{1 / 2} \frac{[x]_{1 / q}-\mu_{q}}{\sigma_{q}}+q^{-1}\right)^{1 / 2}  \tag{2}\\
& \cdot \exp \left(\frac{\log ^{2}\left(q^{-3 / 2}(1-q)^{1 / 2} \frac{[x]_{1 / q}-\mu_{q}}{\sigma_{q}}+q^{-1}\right)}{2 \log q}\right), x \geq 0,
\end{align*}
$$

where $\theta=\theta_{n}, n=0,1,2, \cdots$ such that $\theta_{n}=q^{-\alpha n}$ with $0<a<1$ constant and $\mu_{q}$ and $\sigma_{q}^{2}$ the mean value and variance of the random variable $Y$, respectively. To obtain the above pointwise convergence (2), a $q$ analogue of the well known Stirling formula for the $n$ factorial ( $n!$ ) has been provided.

In statistical mechanics and in computer science such as in probabilistic and approximation algorithms, applications of the $q$-binomial distribution involve sequences of independent Bernoulli trials where in the geometrically decreasing odds of success at the $i$ th trial, the rate $q$ is considered to be a sequence of $n$ with $q=q(n) \rightarrow 1$ as $n \rightarrow \infty$. In this work, under this consideration, a question arises. How this assumption affects the continuous limiting behaviour of this $q$-binomial distribution?

The answer to this question is given in this manuscript by establishing a deformed Gaussian limiting behaviour for the $q(n)$-Binomial distribution is proved. The proofs are concentrated on the study of the sequence $q(n)$ and the parameters of the considered distribution as sequences of $n$. Further, figures using the program MAPLE are presented, indicating the accuracy of the established distribution convergence even for moderate values of $n$.

## 2. Main Results

### 2.1. An Asymptotic Expansion of the $q(n)$-Factorial Number of Order $\boldsymbol{n}$ with $q(n) \rightarrow 1$ as $n \rightarrow \infty$

To initiate our study we need to derive an asymptotic expansion for $n \rightarrow \infty$ of the $q$-factorial number of order n

$$
\begin{align*}
& {[n]_{q}!=[1]_{q}[2]_{q} \cdots[n]_{q}} \\
& =\prod_{k=1}^{n} \frac{1-q^{k}}{(1-q)^{n}}=\frac{(q ; q)_{n}}{(1-q)^{n}}, \tag{3}
\end{align*}
$$

where $q=q(n)$ with $q(n) \rightarrow 1$ as $n \rightarrow \infty$ and
$[t]_{q}=\frac{1-q^{t}}{1-q}$, the $q$-number $t$.
The derived estimate for the $q$-factorial numbers of order $n$, is based on the analysis of the $q$-Exponential function

$$
\begin{align*}
E_{q}((1-q) x) & =\sum_{n=0}^{\infty} \frac{q^{\binom{n}{2}}(1-q)^{n}}{(q ; q)_{n}} x^{n} \\
& =\sum_{n=0}^{\infty} \frac{q^{\binom{n}{2}}}{[n]_{q}!} x^{n}  \tag{4}\\
& =\prod_{j=1}^{\infty}\left(1+(1-q) x q^{j-1}\right)
\end{align*}
$$

which is the ordinary generating function (g.f.) of the numbers $\frac{q^{\binom{n}{2}}}{[n]_{q}!}, n=0,1,2, \cdots$.

Rewriting $E_{q}((1-q) x)$ as follows

$$
\begin{equation*}
E_{q}(x)=\exp (g(x)) \tag{5}
\end{equation*}
$$

where

$$
\begin{equation*}
g(x)=\sum_{j=1}^{\infty} \log \left(1+((1-q) x) q^{j-1}\right) \tag{6}
\end{equation*}
$$

because of the large dominant singularities of the generating function $E_{q}(x)$, a well suited method for analyzing this is the saddle point method.

Using an approach of the saddle point method inspired from [9-12] and [1], the following theorem gives an asymptotic for the $q(n)$-factorial number of order $n$.

Theorem 1. The $q$-factorial numbers of order $n,[n]_{q}$ !, where
A) $q=q(n)$ with $q(n) \rightarrow 1$ as $n \rightarrow \infty$
and $q(n)^{n}=\Omega(1)$
or
B) $q=q(n)$ with $q(n)^{n}=o(1)$
have the following asymptotic expansion for $n \rightarrow \infty$

$$
\begin{align*}
& {[n]_{q}!=(2 \pi)^{1 / 2} q^{\binom{n}{2}}} \\
& \exp (-g(r))\left[r g^{\prime}(r)+r^{2} g^{\prime \prime}(r)\right]^{1 / 2} r^{n} \\
& \cdot\left[1+\sum_{k^{\prime}=1}^{N} S_{k^{\prime}}(r)\left(q^{n}(1-q)\right)^{k^{\prime}}+O\left(\left(q^{n}(1-q)\right)^{\frac{N+1}{2}}\right)\right]^{-1} \tag{7}
\end{align*}
$$

where $N$ is a positive integer, $r$ is the real solution of the equation

$$
r g^{\prime}(r)=n
$$

and

$$
\begin{align*}
& S_{k^{\prime}}(r)=\frac{r^{k^{\prime}}}{\left(2 k^{\prime}\right)!} \\
& \cdot \sum_{j=1}^{2 k^{\prime}} B_{2 k^{\prime}, j}\left(\alpha_{1}(r), \cdots, \alpha_{2 k^{\prime}}(r)\right) \frac{\Gamma\left(j+k^{\prime}+1 / 2\right)}{\pi^{1 / 2}}, \\
& \alpha_{k^{\prime}}(r)=\frac{\left[2^{-1}\left(1+\frac{r g^{\prime \prime}(r)}{g^{\prime}(r)}\right)\right]^{-k^{\prime} / 2-1} \mathrm{i}^{k^{\prime}+2}}{g^{\prime}(r)\left(k^{\prime}+1\right)\left(k^{\prime}+2\right) r^{k^{\prime \prime 2}+1}}  \tag{8}\\
& \cdot \sum_{v=1}^{k^{\prime}+2} S\left(k^{\prime}+2, v\right) r^{v} g^{(v)}(r)
\end{align*}
$$

with $B_{k^{\prime}, j}\left(\psi_{1}, \cdots, \psi_{k^{\prime}}\right)$ the partial Bell polynomials, $S\left(k^{\prime}, j\right)$ the Stirling numbers of the second kind and $i^{2}=-1$.
Proof. We shall study the asymptotic behaviour of the $q$-factorial numbers of order $n,[n]_{q}!$, by expressing them via Cauchy's integral formula that gives the coefficients of a power series:

$$
\begin{equation*}
\frac{q^{n}\binom{n}{2}}{[n]_{q}!}=\frac{1}{2 \pi i} \int_{|x|=r} \frac{\exp (g(x))}{x^{n+1}} \mathrm{~d} x \tag{9}
\end{equation*}
$$

where the contour of integration is taken to be a circle of radius $r$. This integral will be estimated with the saddle point method. The saddle point is defined by the equation $x g^{\prime}(x)=n+1$. It turns out that it is convenient to switch to polar coordinates, setting $x=r e^{i \theta}$. Then the original integral becomes

$$
\begin{align*}
& \frac{\left.q^{n} \begin{array}{l}
n \\
2
\end{array}\right)}{[n]_{q}!}=\frac{\exp (g(r))}{r^{n} 2 \pi}  \tag{10}\\
& \cdot \int_{-\pi}^{\pi} \exp \left[g\left(r \mathrm{re}^{\mathrm{i} \theta}\right)-g(r)-i n \theta\right] \mathrm{d} \theta .
\end{align*}
$$

In accordance with the saddle point method principles, we choose the radius $r$ to be the solution of $r g^{\prime}(r)=n$. Setting $G(\theta)=g\left(r \mathrm{e}^{\mathrm{i} \theta}\right)-g(r)-i n \theta$ with a Maclaurin series expansion about $\theta=0$ we have

$$
\begin{equation*}
G(\theta)=-\phi^{2}+\phi^{2} \sum_{k^{\prime}=1}^{\infty} \alpha_{k^{\prime}}(r) \frac{(\psi \phi)^{k^{\prime}}}{k^{\prime}!} \tag{11}
\end{equation*}
$$

where

$$
\begin{equation*}
\phi=\left[\frac{1}{2}\left(r g^{\prime}(r)+r^{2} g^{\prime \prime}(r)\right)\right]^{1 / 2} \theta, \psi=\left[g^{\prime}(r)\right]^{-\frac{1}{2}} \tag{12}
\end{equation*}
$$

and

$$
\begin{equation*}
\alpha_{k^{\prime}}(r)=\frac{\left.\left[2^{-1}\left(1+\frac{r g^{\prime \prime}(r)}{g^{\prime}(r)}\right)\right]^{-k^{\prime} / 2-1} g^{\left(k^{\prime}+2\right)}\left(r \mathrm{e}^{i \theta}\right)\right|_{\theta=0}}{g^{\prime}(r)\left(k^{\prime}+1\right)\left(k^{\prime}+2\right) r^{k^{\prime} / 2+1}}, \tag{13}
\end{equation*}
$$

where

$$
\left.g^{\left(k^{\prime}+2\right)}\left(r \mathrm{e}^{i \theta}\right)\right|_{\theta=0}=\left.\left(\frac{\mathrm{d}}{\mathrm{~d} \theta}\right)^{k^{\prime}+2} g\left(r \mathrm{e}^{i \theta}\right)\right|_{\theta=0}
$$

The absence of a linear term in $\theta$ indicates a saddle point. The function $\mid e^{G(\theta)}$ is unimodal with its peak at $\theta=0$.

An estimation of the $q$-factorial numbers of order $n,[n]_{q}$ ! with $q$ defined by conditions (A) or (B) should naturally proceed by isolating separately small portions of the contour (corresponding to $x$ near the real axis) as follows.
A) For $q=q(n)$ with $q(n)^{n}=\Omega(1)$ we set

$$
\begin{align*}
& I_{1}=\frac{1}{\sqrt{2 \pi}} \int_{-\delta}^{\delta} \exp [G(\theta)] \mathrm{d} \theta \\
& I_{2}=\frac{1}{\sqrt{2 \pi}} \int_{\delta}^{2 \pi-\delta} \exp [G(\theta)] \mathrm{d} \theta \tag{14}
\end{align*}
$$

and choose $\delta$ such that the following conditions are true (see [12]):

C1) $n \delta^{2} \rightarrow \infty$, that is $\delta \gg n^{-1 / 2}$
C2) $n \delta^{3} \rightarrow 0$, that is $\delta \ll n^{-1 / 3}$,
where "<" means "much smaller than". A suitable choice for $\delta$ is $n^{-3 / 8}$.

As $\left|\mathrm{e}^{G(\theta)}\right|$ decreases in $[\delta, \pi]$,

$$
\begin{equation*}
\left|\mathrm{e}^{G(\theta)}\right| \leq\left|\mathrm{e}^{G(\delta)}\right|, \delta \leq \theta \leq 2 \pi-\delta \tag{15}
\end{equation*}
$$

We will show in the sequel that from C 1 ) and C 2 ) it follows that $\mathrm{e}^{G(\delta)}$ is exponentially small, being dominated by a term of the form $\mathrm{e}^{-o\left(n^{1 / 4}\right)}$.

Indeed we have

$$
G(\delta) \sim-\frac{1}{2}\left(r g^{\prime}(r)+r^{2} g^{\prime \prime}(r)\right) \delta^{2}
$$

or

$$
\begin{equation*}
G(\delta) \sim-\frac{1}{2}\left(r g^{\prime}(r)+r^{2} g^{\prime \prime}(r)\right) n^{-3 / 4} . \tag{16}
\end{equation*}
$$

But

$$
r g^{\prime}(r)+r^{2} g^{\prime \prime}(r) \sim \frac{1}{\log q^{-1}}\left(q^{-n}-1\right)
$$

or

$$
\begin{equation*}
r g^{\prime}(r)+r^{2} g^{\prime \prime}(r) \sim \frac{(1-q)}{\log q^{-1}}\left(q^{n-1}+q^{n-2}+\cdots+1\right) \tag{17}
\end{equation*}
$$

For $q=q(n)$ with $q(n)^{n}=\Omega(1)$ we get

$$
\begin{equation*}
G(\delta)=O\left(-\frac{1}{2} n^{1 / 4}\right) \tag{18}
\end{equation*}
$$

From which we find that

$$
\begin{equation*}
\left|I_{2}\right|=O\left(\mathrm{e}^{G(\delta)}\right)=O\left(\mathrm{e}^{-\frac{1}{2} n^{\prime / 4}}\right) . \tag{19}
\end{equation*}
$$

Thus, by C 1$), \delta$ has been taken large enough so that the central integral $I_{1}$ "captures" most of the contribution, while the remainder integral $I_{2}$ is exponentially small by (19).

We now turn to the precise evaluation of the central integral $I_{1}$. We have

$$
\begin{align*}
I_{1}= & \frac{1}{\left[r g^{\prime}(r)+r^{2} g^{\prime \prime}(r)\right]^{1 / 2}} \frac{1}{\sqrt{\pi}}  \tag{20}\\
& \cdot \int_{-\epsilon}^{\epsilon} \exp \left[-\phi^{2}+\phi^{2} \sum_{k^{\prime}=1}^{\infty} \alpha_{k^{\prime}}(r) \frac{(\psi \phi)^{k^{\prime}}}{k^{\prime}!}\right] \mathrm{d} \phi
\end{align*}
$$

where

$$
\begin{equation*}
\epsilon=\left[\frac{1}{2}\left(r g^{\prime}(r)+r^{2} g^{\prime \prime}(r)\right)\right]^{1 / 2} \delta . \tag{21}
\end{equation*}
$$

Note that $\epsilon \rightarrow \infty$ as $n \rightarrow \infty$, since

$$
\begin{aligned}
\epsilon & =n^{-3 / 8}\left[\frac{1}{2}\left(r g^{\prime}(r)+r^{2} g^{\prime \prime}(r)\right)\right]^{1 / 2} \\
& =n^{1 / 8}\left[\frac{1}{2}\left(1+\frac{r g^{\prime \prime}(r)}{g^{\prime}(r)}\right)\right]^{1 / 2}>C n^{1 / 8},
\end{aligned}
$$

where $C$ a positive constant.
B) For $q=q(n)$ with $q(n)^{n}=o(1)$ we set

$$
\begin{align*}
& I_{3}=\frac{1}{\sqrt{2 \pi}} \int_{-\delta}^{\delta} r^{-n} \exp [G(\theta)] \mathrm{d} \theta  \tag{22}\\
& I_{4}=\frac{1}{\sqrt{2 \pi}} \int_{\delta}^{2 \pi-\delta} r^{-n} \exp [G(\theta)] \mathrm{d} \theta
\end{align*}
$$

and choose $\delta$ such that the conditions C 1 ) and C 2 ) are true. We suitably select $\delta=n^{-3 / 8}$.

As $\left|\mathrm{e}^{G(\theta)}\right|$ decreases in $[\delta, \pi]$,

$$
\begin{equation*}
\left|e^{G(\theta)}\right| \leq\left|\mathrm{e}^{G(\delta)}\right|, \delta \leq \theta \leq 2 \pi-\delta \tag{23}
\end{equation*}
$$

We will now show that $\mathrm{e}^{G(\delta)}$ is dominated by a term of the form $O(1)$. Indeed, form C 1$), \mathrm{C} 2), 16$ ) and 17) it follows that

$$
\begin{equation*}
\exp (G(\delta)) \sim \exp \left(-\frac{1}{2 \log q^{-1}}\left(1-q^{n}\right) n^{-3 / 4}\right) \tag{24}
\end{equation*}
$$

From which we get

$$
\begin{equation*}
\left|I_{4}\right|=O\left(r^{-n} \mathrm{e}^{G(\delta)}\right)=O\left(q^{n^{2}} \mathrm{e}^{-\frac{1}{2} \frac{1}{\log q^{-1}}\left(1-q^{n}\right)^{n^{-3 / 4}}}\right) \tag{25}
\end{equation*}
$$

Thus, for $q=q(n)$ with $q(n)^{n}=o(1)$ the integral $I_{4}$ is negligibly small. We now turn to the precise evaluation of the central integral $I_{3}$. Since

$$
\begin{aligned}
I_{3}= & \frac{1}{\sqrt{2 \pi}} \int_{-\delta}^{\delta} r^{-n} \exp [G(\theta)] \mathrm{d} \theta \\
& =\frac{1}{\sqrt{2 \pi}} \int_{-\infty}^{\infty} r^{-n} \exp [G(\theta)] \mathrm{d} \theta \\
& -\frac{1}{\sqrt{2 \pi}} \int_{-\infty}^{-\delta} r^{-n} \exp [G(\theta)] \mathrm{d} \theta \\
& -\frac{1}{\sqrt{2 \pi}} \int_{\delta}^{\infty} r^{-n} \exp [G(\theta)] \mathrm{d} \theta \\
= & \frac{1}{\sqrt{2 \pi}} \int_{-\infty}^{\infty} r^{-n} \exp [G(\theta)] \mathrm{d} \theta \\
& -O\left(2 q^{n^{2}} \mathrm{e}^{-\frac{1}{2} \frac{1}{\log q^{-1}}\left(1-q^{n}\right) n^{-3 / 4}}\right)
\end{aligned}
$$

we have

$$
\begin{align*}
I_{3}= & \frac{r^{-n}}{\left[r g^{\prime}(r)+r^{2} g^{\prime \prime}(r)\right]^{1 / 2}} \frac{1}{\sqrt{\pi}} \\
& \cdot \int_{-\infty}^{\infty} \exp \left[-\phi^{2}+\phi^{2} \sum_{k=1}^{\infty} \alpha_{k}(r) \frac{(\psi \phi)^{k}}{k!}\right] \mathrm{d} \phi . \tag{26}
\end{align*}
$$

We now unifiable proceed our proof for both conditions A) and B) and working analogously as in Kyriakoussis and Vamvakari [1] we get our final estimation (7). $\diamond$

In the previous theorem due to saddle point method principles, we have chosen the radius $r$ of the derived asymptotic expansion (7) to be the solution of $r g^{\prime}(r)=n$. By solving this saddle point equation we get that

$$
r=q^{-1}[n]_{1 / q}
$$

and

$$
g^{\prime}(r)+r^{2} g^{\prime \prime}(r) \cong \frac{1-q}{\log q^{-1}}[n]_{1 / q} q^{n-1}
$$

So, by substituting these to our estimation (7) the following corollary is proved.

Corollary 1. The $q$-factorial numbers of order $n,[n]_{q}!$, where
A) $q=q(n)$ with $q(n) \rightarrow 1$ as $n \rightarrow \infty$ and $q(n)^{n}=\Omega(1)$
or
B) $q=q(n)$ with $q(n)^{n}=o(1)$
have the following asymptotic expansion for $n \rightarrow \infty$

$$
\begin{align*}
{[n]_{q}!=} & \frac{(2 \pi(1-q))^{1 / 2}}{\left(q \log q^{-1}\right)^{1 / 2}} \\
& \cdot \frac{q^{\binom{n}{2}} q^{n / 2}[n]_{1 / q}^{n+1 / 2}}{\prod_{j=1}^{\infty}\left(1+\left(q^{-n}-1\right) q^{j-1}\right)}\left(1+O\left(q^{n}(1-q)\right)\right) \tag{27}
\end{align*}
$$

### 2.2. Deformed Gaussian Limiting Behaviour for the $\boldsymbol{q}(\boldsymbol{n})$-Binomial Distributions with $\boldsymbol{q}(\boldsymbol{n}) \rightarrow \mathbf{1}$ as $\boldsymbol{n} \rightarrow \infty$

Transferred from the random variable $X$ of the $q$ binomial distribution (1) to the equal-distributed deformed random variable $Y=[X]_{1 / q}$, the mean value and variance of the random variable $Y$, say $\mu_{q}$ and $\sigma_{q}^{2}$ respectively, are given by the next relations

$$
\begin{equation*}
\mu_{q}=[n]_{q} \frac{\theta}{1+\theta q^{n-1}} \tag{28}
\end{equation*}
$$

and

$$
\begin{align*}
\sigma_{q}^{2}= & \frac{\theta^{2}[n]_{q}[n-1]_{q}}{q\left(1+\theta q^{n-1}\right)\left(1+\theta q^{n-2}\right)} \\
& +\frac{\theta[n]_{q}}{\left(1+\theta q^{n-1}\right)}-\frac{\theta^{2}[n]_{q}^{2}}{\left(1+\theta q^{n-1}\right)^{2}} \tag{29}
\end{align*}
$$

(see Kyriakoussis and Vamvakari [1]).
Using the standardized r.v.

$$
Z=\frac{[X]_{1 / q}-\mu_{q}}{\sigma_{q}}
$$

with $\mu_{q}$ and $\sigma_{q}$ given in (28) and (29), the $q$-analogue Stirling asymptotic formula (27) and inspired by [1], the following theorem explores the continuous limiting behaviour of the $q(n)$-binomial distribution with $q(n) \rightarrow 1$ as $n \rightarrow \infty$.
Theorem 2. Let the p.f. of the q-binomial distribution be of the form

$$
f_{X}(x)=\binom{n}{x}_{q} q^{\binom{x}{2}} \theta^{x} \prod_{j=1}^{n}\left(1+\theta q^{j-1}\right)^{-1}, x=0,1, \cdots, n,
$$

where $\theta=\theta_{n}, n=0,1,2, \cdots$ such that $\theta_{n} \rightarrow \infty$, as $n \rightarrow \infty$. Then, for

$$
\text { A) } q=q(n) \text { with } q(n)^{n}=\Omega(1)
$$

or

$$
\text { B) } q=q(n) \text { with } q(n) \rightarrow 1 \text { as } n \rightarrow \infty \text { and } q(n)^{n}=o(1)
$$

and $\theta_{n}=q^{-\alpha n}$ with $0<a<1$ constant
the $q(n)$-binomial distribution is approximated, for $n \rightarrow \infty$, by a deformed standardized Gauss distribution as follows

$$
\begin{align*}
& f_{X}(x) \cong \frac{\left(\log q^{-1}\right)^{1 / 2}}{(2 \pi)^{1 / 2}} \\
& \cdot \exp \left(-\frac{1}{2}\left(\frac{\sigma_{q}}{\mu_{q}\left(\log q^{-1}\right)^{1 / 2}} \frac{[x]_{1 / q}-\mu_{q}}{\sigma_{q}}\right)^{2}\right), x \geq 0 . \tag{30}
\end{align*}
$$

Proof. Using the $q$-analogue of Stirling type (27), for $q=q(n)$ with $q(n) \rightarrow 1$ and $q(n)^{n}=\Omega(1)$ or $q(n)^{n}=o(1)$, the $q$-binomial distribution (1), is approximated by

$$
\begin{align*}
f_{X}(x) \cong & \frac{\left(q \log q^{-1}\right)^{1 / 2}}{(2 \pi(1-q))^{1 / 2}} \frac{\theta_{n}^{x}}{(1-q)^{x}} \\
& \cdot \frac{\prod_{j=1}^{\infty}\left(1+\left(q^{-x}-1\right) q^{j-1}\right)}{\prod_{j=1}^{\infty}\left(1+\theta_{n} q^{j-1}\right) q^{x / 2}[x]_{1 / q}^{x+1 / 2}} . \tag{31}
\end{align*}
$$

Let the random variable $[X]_{\frac{1}{q}}=\frac{1-q^{-X}}{1-q^{-1}}$ and the $q$ standardized r.v. $Z=\frac{[X]_{1 / q}-\mu_{q}}{\sigma_{q}}$ with $\mu_{q}$ and $\sigma_{q}$ given by (28) and (29) respectively, then all the following listed estimations are easily derived

$$
\begin{gather*}
{[x]_{1 / q} \cong \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right),}  \tag{32}\\
q^{-x} \cong\left(q^{-1}-1\right) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right)+1,  \tag{33}\\
x \cong \frac{1}{\log q^{-1}} \log \left(\left(q^{-1}-1\right) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right)+1\right), \tag{34}
\end{gather*}
$$

$$
\begin{align*}
& \left([x]_{1 / q}\right)^{x} \\
& \cong \mu_{q}^{x} \cdot \exp \left(\frac{1}{\log q^{-1}} \log \left(\left(q^{-1}-1\right) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right)+1\right)\right.  \tag{35}\\
& \left.\quad \cdot \log \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right)\right)
\end{align*}
$$

Also, the estimation of the next product

$$
\begin{align*}
& \prod_{j=1}^{\infty}\left(1+q\left(q^{-x}-1\right) q^{j-1}\right) \\
= & \prod_{j=1}^{\infty}\left(1+q\left(q^{-1}-1\right) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right) q^{j-1}\right)  \tag{36}\\
= & \exp \left(\sum_{j=1}^{\infty} \log \left(1+(1-q) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right) q^{j-1}\right)\right),
\end{align*}
$$

is derived by applying the Euler-Maclaurin summation formula (see Odlyzko [8], p. 1090) in the sum of the above Equation (36) as follows

$$
\begin{align*}
& \sum_{j=1}^{\infty} \log \left(1+(1-q) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right) q^{j-1}\right) \\
= & \frac{1}{2 \log q^{-1}} \log ^{2}\left((1-q) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right)\right) \\
+ & +i_{2}\left(\frac{(1-q) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right)}{\left.(1-q) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right)+1\right)}\right.  \tag{37}\\
+ & \frac{1}{2} \log \left(1+(1-q) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right)\right) \\
+ & \frac{\beta_{2} \log q}{2} \underbrace{\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right)}_{1+(1-q) \mu_{q}\left(\frac{\sigma_{q}}{\mu_{q}} z+1\right)}+O(\log q),
\end{align*}
$$

where $L i_{2}$ the dilogarithmic function and $\beta_{2}$ the Bernoulli number of order 2 .

Moreover, working similarly for the sum appearing in the product

$$
\begin{equation*}
\prod_{j=1}^{\infty}\left(1+\theta_{n} q^{j-1}\right)=\exp \left(\sum_{j=1}^{\infty} \log \left(1+\theta_{n} q^{j-1}\right)\right) \tag{38}
\end{equation*}
$$

the next estimation is obtained

$$
\begin{align*}
\sum_{j=1}^{\infty} \log \left(1+\theta_{n} q^{j-1}\right)= & \frac{1}{2 \log q^{-1}} \log ^{2}\left(\theta_{n}\right) \\
& +L i_{2}\left(\frac{\theta_{n}}{\theta_{n}+1}\right)+\frac{1}{2} \log \left(1+\theta_{n}\right)  \tag{39}\\
& +\frac{\beta_{2} \log q}{2} \frac{\theta_{n}}{1+\theta_{n}}+O\left(\theta_{n}^{-1}\right)
\end{align*}
$$

Applying all the previous the estimations (32)-(39) to the approximation (31), carrying out all the necessary manipulations and for $\theta_{n} \rightarrow \infty$, by both conditions A) and B), we derive our final asymptotic (30). $\diamond$

Remark 2. A realization of the sequence $q(n), n=0,1,2, \cdots$ considered in the above theorem 1A) is

$$
q(n)=1-\frac{\beta}{n}, 0<\beta \leq 1
$$

with

$$
q(n)^{n}=\exp (-\beta)
$$

Remark 3. Possible realizations of the sequence $q(n), n=0,1,2, \cdots$ considered in the above theorem 2B) are among others the next two ones

$$
q(n)=1-\frac{1}{\ln (n)} \text { or } q(n)=1-\frac{1}{n^{c}}, 0<c<1 .
$$

Corollary 2 Let the random variable $X$ with p.f. that of the $q(n)$-binomial distribution as in Theorem 2. Then for $n \rightarrow \infty$ the following approximation holds

$$
\begin{align*}
& P(a \leq X \leq b) \cong \frac{1}{2} \operatorname{Erf}\left(u_{b+1}\right)-\frac{1}{2} \operatorname{Erf}\left(u_{a}\right),  \tag{40}\\
& 0 \leq a \leq b,
\end{align*}
$$

where

$$
\begin{equation*}
u_{a}=\frac{\left\{[a-1 / 2]_{1 / q}-\mu_{q}\right\} / \sigma_{q}}{\mu_{q}\left(2 \log q^{-1}\right)^{1 / 2}} \tag{41}
\end{equation*}
$$

with $\operatorname{Erf}(t)=\frac{2}{\sqrt{\Pi}} \int_{0}^{t} \exp \left(-x^{2}\right) \mathrm{d} x$ the Gauss error function.

Proof. Using the approximation (2) and the classical continuity correction we have that

$$
\begin{align*}
& P[a \leq X \leq b]=\sum_{x \in[a, b]} P(X=x) \\
& \cong \int_{a-\frac{1}{2}}^{b+\frac{1}{2}} \frac{\left(\log q^{-1}\right)^{1 / 2}}{(2 \pi)^{1 / 2}}  \tag{42}\\
& \cdot \exp \left(-\frac{1}{2}\left(\frac{\sigma_{q}}{\mu_{q}\left(\log q^{-1}\right)^{1 / 2}} \frac{[x]_{1 / q}-\mu_{q}}{\sigma_{q}}\right)^{2}\right) \mathrm{d} x .
\end{align*}
$$

Setting

$$
Z=\frac{[X]_{1 / q}-\mu_{q}}{\sigma_{q}}
$$

the approximation (42) becomes

$$
\begin{align*}
& P[a \leq X \leq b]=\sum_{x \in[a, b]} P(X=x) \\
& \cong \frac{\sigma_{q} \mu_{q}^{-1}}{\left(\log q^{-1}\right)^{1 / 2}(2 \pi)^{1 / 2}}  \tag{43}\\
& \cdot \int_{\frac{[a-1 / 2]_{q / q}-\mu_{q}}{\sigma_{q}}}^{\frac{[b+1 / 2]_{1 / q}-\mu_{q}}{}} \exp \left(-\frac{1}{2}\left(\frac{\sigma_{q} z}{\mu_{q}\left(\log q^{-1}\right)^{1 / 2}}\right)^{2}\right) \mathrm{d} z .
\end{align*}
$$

Carrying out all the necessary manipulations, we get the final approximation (40). $\diamond$

## 3. Figures Using Maple

In this section, we present a computer realization of approximation (30), by providing figures using the computer program MAPLE and the $q$-series package developed by F. Garvan [13] which indicate good convergence even
for moderate values of $n$. Analytically, for the random variable $X$, we give the Figures 1 and 2 realizing Theorem 2(A), by demonstrating with diamond blue points the exact probability

$$
\begin{align*}
& f_{X}(x)=P(X=x)=\operatorname{Prob}\left(x-\frac{1}{2} \leq X \leq x+\frac{1}{2}\right)  \tag{44}\\
& x=0,1,2, \cdots, n
\end{align*}
$$

and with diamond green points the continuous probability approximation

$$
\begin{align*}
& b_{n}^{q}(x)=\operatorname{Prob}\left(x-\frac{1}{2} \leq X \leq x+\frac{1}{2}\right) \\
& \cong \frac{1}{2} \operatorname{Erf}\left(u_{x+1}\right)-\frac{1}{2} \operatorname{Erf}\left(u_{x}\right),  \tag{45}\\
& 0 \leq x \leq n
\end{align*}
$$

with $u_{x}$ and $u_{x+1}$ given by Equation (41), for $q=q(n)=1-\frac{1}{n}$,
$\theta=\theta_{n}=(\exp (1)-1) /\left(\exp (1)-2 \exp (1) q^{n}+q^{n}\right)$
and $n=50,100$.
Note that similar good convergence even for moderate values of $n$ have been implemented for Theorem 2B).

The procedure in MAPLE which realizes the exact probability (44) and its approximation (45) for given $n, q$ and theta for both Theorem 2 A ) and 2 B ), is available under request.

## 4. Concluding Remarks

In this article, a deformed Gaussian limiting behaviour


Figure 1. Sketch of exact probability (44) by blue diamond points and probability approximation (45) by green diamond points, for $\boldsymbol{n}=\mathbf{5 0}$.


Figure 2. Sketch of exact probability (44) by blue diamond points and probability approximation (45) by green diamond points, for $\boldsymbol{n}=100$.
for the $q(n)$-Binomial distribution has been established. The proofs have been concentrated on the study of the sequence $q(n)$ and the parameters of the considered distributions as sequences of $n$. Further, figures using the program MAPLE have been presented, indicating the accuracy of the established distribution convergence even for moderate values of $n$.

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