# Approximating the negative moments of the Poisson distribution 

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#### Abstract

Through the use of the Stirling numbers of the first kind, a general class of approximations is derived for the negative moments of the Poisson distribution.


Keywords: Negative moments; Stirling numbers of the first kind; Poisson distribution

## 1 Introduction

The approximation of negative moments of the Poisson distribution is an important practical problem and has attracted significant attention in literature, see for example Chao and Strawderman (1972), Gupta (1979), Stancu (1968), Stephan (1945) and Tiku (1964). The main applications of the first negative moment are related to the theory of mixed Poisson distributions: indeed, if $\eta_{i}(i=1,2, \ldots n)$ are i.i.d.r.v. with variance $\sigma^{2}$ and the sample size $n$ has the positive Poisson distribution, then the variance of the sample mean $\frac{1}{n} \sum_{i=1}^{n} \eta_{i}$ is $\sigma^{2} \mathrm{E}(1 / n)$, see e.g. Grab and Savage (1954). The negative moments of the Poisson distribution are known to be useful in life testing problems, as shown in Bartholomew (1957), David and Johnson (1956) and Epstein and Sobel (1954).

A recent application has occurred in the field of multi-centre clinical trials, which involves the simultaneous recruitment of patients to numerous centres. The approximation of negative moments are required in the subsequent calculations of MSE's of the mean treatment effect differences. As demonstrated in Dragalin et al. (2002), three least squared estimators can be defined for the mean treatment difference based on fixed effect models of increasing

[^0]complexity. In particular, the $M S E$ of the Type III estimator is
\[

$$
\begin{equation*}
\frac{\sigma^{2}}{N^{2}} \sum_{i=1}^{N}\left(\frac{1}{n_{i 2}}+\frac{1}{n_{i 1}}\right) \tag{1}
\end{equation*}
$$

\]

where $N$ is the number of centres, $\sigma^{2}$ is the variance of the response observed at each centre, and $n_{i j}$ is the number of patients of treatment $j$ at centre $i$. If $n_{i j}$ are assumed to be positive Poisson random variables then (1) becomes a random variable, with moments that are expressed through the negative moments of the Poisson distribution. Similar phenomena occurs with the Type I and II estimators, however, with a more complicated form. See Fedorov et al. (2003) for further details.

Let $\operatorname{Poisson}(\lambda)$ denote the Poisson distribution with parameter $\lambda$, and let $\xi$ be a random variable, $\xi \sim \operatorname{Poisson}(\lambda)$. Additionally, let $\xi_{+}$be the so-called positive Poisson r.v. with parameter $\lambda$; that is,

$$
\operatorname{Pr}\left(\xi_{+}=x\right)=\frac{1}{1-e^{-\lambda}} \frac{\lambda^{x}}{x!} e^{-\lambda}, \text { for } x=1,2 \ldots
$$

The negative moments of the Poisson distribution are defined as the negative moments of $\xi_{+}$:

$$
\begin{equation*}
\mu_{-\alpha}=E\left(\frac{1}{\xi_{+}^{\alpha}}\right)=\frac{e^{-\lambda}}{1-e^{-\lambda}} \sum_{x=1}^{\infty} \frac{\lambda^{x}}{x^{\alpha} x!} \text { for any } \alpha>0 . \tag{2}
\end{equation*}
$$

We develop the following $k^{\text {th }}$ order approximations for $\mu_{-\alpha}$, with integer $\alpha$ :

$$
\begin{equation*}
\mu_{-\alpha} \cong \mu_{-\alpha}^{(k)}=\sum_{u=\alpha}^{k} \frac{s(u, \alpha)}{\lambda^{u}}, \tag{3}
\end{equation*}
$$

where $s(u, \alpha)$ are the so-called signless Stirling numbers of the first kind, see Section 3.1. The case of non-integer $\alpha$ is considered in Section 3.6.

Approximations of the negative moments $\mu_{-\alpha}$ introduced in Tiku (1964) have been cited in a number of reference books; see, for instance, Haight (1966), Johnson et al. (1992). We shall use these approximations as a benchmark for comparison in subsequent calculations, and we will demonstrate that our approximations are more precise. Tiku's approximations for the first negative moment (see Equation (11) in Tiku (1964)) are defined as follows:

$$
\begin{equation*}
\mu_{-1} \cong T_{j}^{(1)}=\frac{\left(1+\sum_{r=3}^{j} \beta_{r}\right)}{(\lambda-1)\left(1-e^{-\lambda}\right)} ; \text { with } T_{0}^{(1)}=\frac{1}{(\lambda-1)\left(1-e^{-\lambda}\right)}, \tag{4}
\end{equation*}
$$

where $\beta_{r}=a^{(r)} / \lambda(\lambda+1)(\lambda+2) \ldots(\lambda+r-1), r=3,4,5, \ldots$ with $a^{(3)}=1, a^{(4)}=7, a^{(5)}=$ $43, a^{(6)}=271$, etc.

Note that a number of alternative approximations to $\mu_{-1}$ can be constructed through using the fact that

$$
\begin{equation*}
\mu_{-1}=\frac{e^{-\lambda}}{1-e^{-\lambda}}(\operatorname{Ei}(\lambda)-\log \lambda-\gamma), \tag{5}
\end{equation*}
$$

where $\gamma \simeq 0.5772 \ldots$ is Euler's constant and $\mathrm{E}_{i}(\lambda)=\int_{-\infty}^{\lambda}\left(e^{u} / u\right) d u$ is the exponential integral, see for example Grab and Savage (1954). Note that the relation (5) is often used to approximate values of the exponential integral $\operatorname{Ei}(\lambda)$; consequently, this broadens the scope of our approximation into several other fields.

The approximations of Tiku (1964) for the higher order integer negative moments ( $\alpha \geq 2$ ) are defined as

$$
\begin{equation*}
\mu_{-\alpha} \cong T^{(\alpha)}=\frac{1}{(\lambda-1)(\lambda-2) \ldots(\lambda-\alpha)} . \tag{6}
\end{equation*}
$$

The suggested approximations are far more precise than Tiku's approximations. The criterion by which we determine the accuracy of the approximations is through studying their relative error:

$$
\text { Relative Error }=\frac{\text { Exact Value }- \text { Approximate Value }}{\text { Exact Value }} .
$$

Note that positive values for the relative error indicate that the respective approximation is underestimating the true value.

In this paper we are only interested in moderate and large values of $\lambda$. For small $\lambda$ the moments can be easily computed using the definition (2). In particular, the relative error of the following simple approximation

$$
\mu_{-\alpha}=\frac{e^{-\lambda}}{1-e^{-\lambda}} \sum_{k=1}^{\infty} \frac{\lambda^{k}}{k^{\alpha} k!} \cong \frac{e^{-\lambda}}{1-e^{-\lambda}} \sum_{k=1}^{3 \lambda+10} \frac{\lambda^{k}}{k^{\alpha} k!} ;
$$

is smaller than $10^{-10}$ in absolute value for all $\lambda>0$ and $\alpha \geq 1$.

## 2 Deriving recurrence relations and approximations

### 2.1 General Recurrences

Denote

$$
\begin{equation*}
A_{l, m}(\lambda)=\sum_{x=1}^{\infty} \frac{e^{-\lambda} \lambda^{x}}{x^{l}(x+m-1)!} . \tag{7}
\end{equation*}
$$

In particular,

$$
\begin{equation*}
\mu_{-\alpha}=\frac{1}{1-e^{-\lambda}} A_{\alpha, 1}(\lambda) . \tag{8}
\end{equation*}
$$

For each $l, m>0$, we obviously have

$$
\begin{equation*}
\frac{1}{x^{l}}=\frac{1}{x^{l-1}(x+m)}+\frac{m}{x^{l}(x+m)} \tag{9}
\end{equation*}
$$

Through applying Equation (9) to (7) we obtain the basic recurrence:

$$
\begin{equation*}
A_{l, m}(\lambda)=\sum_{x=1}^{\infty} \frac{e^{-\lambda} \lambda^{x}}{x^{l-1}(x+m)!}+\sum_{x=1}^{\infty} \frac{m e^{-\lambda} \lambda^{x}}{x^{l}(x+m)!}=A_{l-1, m+1}(\lambda)+m A_{l, m+1}(\lambda), \tag{10}
\end{equation*}
$$

and from (7) we have

$$
\begin{equation*}
A_{0, m}(\lambda)=\frac{1}{\lambda^{m-1}}\left(1-\sum_{x=0}^{m-1} \frac{e^{-\lambda} \lambda^{x}}{x!}\right) . \tag{11}
\end{equation*}
$$

Note that for any fixed $m \geq 1$ and $\lambda \geq 1$ we have

$$
\begin{equation*}
\frac{1}{\lambda^{m-1}}\left(1-m \lambda^{m-1} e^{-\lambda}\right) \leq A_{0, m}(\lambda) \leq \frac{1}{\lambda^{m-1}} \tag{12}
\end{equation*}
$$

so that the difference $A_{0, m}(\lambda)-1 / \lambda^{m-1}$ is exponentially small as $\lambda \rightarrow \infty$.
Through performing several iterations of the basic recurrence (10), we obtain

$$
\begin{align*}
A_{l, m}(\lambda) & =A_{l-1, m+1}(\lambda)+m A_{l-1, m+2}(\lambda)+m(m+1) A_{l, m+2}(\lambda)=\ldots \\
& =\sum_{r=0}^{N} \frac{(m+r-1)!}{(m-1)!} A_{l-1, m+r+1}(\lambda)+\frac{(m+N)!}{(m-1)!} A_{l, m+N+1}(\lambda) \tag{13}
\end{align*}
$$

Rather than using the definition (7), we can define $A_{l, m}(\lambda)$ through the recurrences (13) and the initial conditions (11). Let the quantities $B_{l, m}(\lambda)$, which can be considered as a simplified version of $A_{l, m}(\lambda)$, be defined through the recurrences (13), that is,

$$
\begin{equation*}
B_{l, m}(\lambda)=\sum_{r=0}^{N} \frac{(m+r-1)!}{(m-1)!} B_{l-1, m+r+1}(\lambda)+\frac{(m+N)!}{(m-1)!} B_{l, m+N+1}(\lambda) \quad \text { for all } N>0, \tag{14}
\end{equation*}
$$

and the initial conditions that are similar to (11) but exclude the exponential terms:

$$
\begin{equation*}
B_{0, m}(\lambda)=\frac{1}{\lambda^{m-1}} . \tag{15}
\end{equation*}
$$

Deriving approximations using $B_{l, m}(\lambda)$ as opposed to $A_{l, m}(\lambda)$ implies that we ignore all the terms that are exponentially decreasing when $\lambda \rightarrow \infty$. Since $B_{0, m}(\lambda)=\frac{1}{\lambda^{m-1}} \geq A_{0, m}(\lambda)$ for all $m$, see (12), we have $B_{l, m}(\lambda) \geq A_{l, m}(\lambda)$ for all $l \geq 0, m \geq 1$.

### 2.2 Deriving approximations and determining their precision

Let $m, l$ and $k$ be fixed. Consider the expansions of $A_{l, m}(\lambda)$ and $B_{l, m}(\lambda)$ of the form:

$$
\begin{align*}
& A_{l, m}(\lambda)=\sum_{r=0}^{k} \frac{a_{l, m}(r)}{\lambda^{r}}+o\left(\frac{1}{\lambda^{k}}\right), \lambda \rightarrow \infty ;  \tag{16}\\
& B_{l, m}(\lambda)=\sum_{r=0}^{k} \frac{b_{l, m}(r)}{\lambda^{r}}+o\left(\frac{1}{\lambda^{k}}\right), \lambda \rightarrow \infty, \tag{17}
\end{align*}
$$

where $a_{l, m}(r)$ and $b_{l, m}(r)$ are some coefficients. The properties of these coefficients are summarized in the following statement.

Theorem 1. Let l, $k \geq 0, m \geq 1$ be some integers and let $A_{l, m}(\lambda)$ and $B_{l, m}(\lambda)$ be defined through (16) and (17), respectively. Then we have
(i) $a_{l, m}(r)=b_{l, m}(r)$ for all $l, m, r$;
(ii) $a_{l, m}(r)=0$ for all $r<l+m-1$;
(iii) $a_{l, m}(r)=1$ for $r=l+m-1$;
(iv) $a_{l, m}(r) \geq 0$ for all $r$.

Proof. The statement (i) follows from the fact that for fixed $l$ and $m$ the difference between $A_{l, m}(\lambda)$ and $B_{l, m}(\lambda)$ is exponentially small as $\lambda \rightarrow \infty$.

In order to prove the other three statements, we use induction in $l$. For $l=0$ the statements obviously follow from either (12) or (15). Assume that (ii) - (iv) hold for $l-1$. For all fixed $m \geq 1, l \geq 1$ we have for all $x \geq 1$ :

$$
\frac{1}{(x+m) \ldots(x+m+l-1)} \leq \frac{1}{x^{l}} \leq \frac{1}{(x+m) \ldots(x+m+l-1)}\left(1+\frac{C_{m, l}}{x+m+l}\right)
$$

for some positive constants $C_{m, l}$; these constants are $C_{m, l}=(m+l+1)((m+l)!/ m!-1)$. Through substituting this into (7) we obtain

$$
A_{0, m+l}(\lambda) \leq A_{l, m}(\lambda) \leq A_{0, m+l}(\lambda)+C_{m, l} A_{0, m+l+1}(\lambda) .
$$

In view of (12) we obtain as $\lambda \rightarrow \infty$

$$
\frac{1}{\lambda^{m+l-1}}+o\left(\frac{1}{\lambda^{m+l}}\right) \leq A_{l, m}(\lambda) \leq \frac{1}{\lambda^{m+l-1}}+\frac{C_{m, l}}{\lambda^{m+l}} ;
$$

this yields (ii) and (iii).
Assume that (iv) is true for $l-1$, through the recurrence relation in (13) it follows that (iv) is true for $l$; as it implies that each $a_{l, m}(r)$ is a weighted sum of non-negative coefficients $a_{l-1, u}(v)$ with non-negative weights.

In view of (8) and (16), we define the $k^{\text {th }}$ order approximations to $\mu_{-l}$ as

$$
\begin{equation*}
\mu_{-l}^{(k)}=\sum_{r=0}^{k} \frac{a_{l, 1}(r)}{\lambda^{r}} . \tag{18}
\end{equation*}
$$

The statements (ii) and (iii) of Theorem 1 imply the summation in (18) actually starts at $r=l$, they also imply that as $\lambda \rightarrow \infty$ the order of decrease of $\mu_{-l}$ is $\frac{1}{\lambda}$; more precisely:

$$
\begin{equation*}
\mu_{-l}-\frac{1}{\lambda^{l}}=o\left(\frac{1}{\lambda^{l}}\right) \text { as } \lambda \rightarrow \infty . \tag{19}
\end{equation*}
$$

Statement (iv) yields that for each $l, \mu_{-l}$ is monotonously increasing as $k$ grows, that is $\mu_{-l}^{(k)} \leq \mu_{-l}^{(k+1)}$ for all $k$.

Assume that we able to compute the coefficients $a_{l, 1}(r)$ (which we can, unlike the coefficients $a_{l, m}(r)$ for $\left.m>1\right)$. Thus, we can compute the $k^{\text {th }}$ order approximations $\mu_{-l}^{(k)}$ to $\mu_{-l}$ with $k>l$, with Theorem 1 implying that the precision is:

$$
\begin{equation*}
\left|\mu_{-l}^{(k)}-\mu_{-l}\right|=O\left(\frac{1}{\lambda^{k+1}}\right) \text { as } \lambda \rightarrow \infty . \tag{20}
\end{equation*}
$$

Formulae (19) and (20) yield that the relative error of $\mu_{-l}^{(k)}$ with $k>l$ is:

$$
\begin{equation*}
\left|\frac{\mu_{-l}^{(k)}-\mu_{-l}}{\mu_{-l}}\right|=\frac{O\left(\frac{1}{\lambda^{k+1}}\right)}{\frac{1}{\lambda^{l}}+o\left(\frac{1}{\lambda^{l}}\right)}=\frac{O\left(\frac{1}{\lambda^{k-l+1}}\right)}{1+o(1)}=O\left(\frac{1}{\lambda^{k-l+1}}\right) \text { as } \lambda \rightarrow \infty . \tag{21}
\end{equation*}
$$

Despite the rate of decrease ( as $\lambda \rightarrow \infty$ ) of the relative error, $\mu_{-l}^{(k)}$ increases as $k$ grows, we cannot take $k$ arbitrarily large. This is attributable to the constant in $O\left(\frac{1}{\lambda^{k-l+1}}\right)$ in (21) rapidly increasing as $k$ grows. This constant, as follows from (2) has the order of at least $s(k+1, l)$, which is roughly $k!(\log k)$. Numerical results (shown below) demonstrate that one can guarantee extremely precise approximations by taking $k$ up to $\lceil\lambda\rceil+l$.

Note also that Theorem 1 implies that to construct a $k^{t h}$ order approximation to $\mu_{-l}$, we are free to ignore the last terms in (14) and use

$$
\begin{equation*}
B_{l, m}(\lambda) \cong \sum_{r=0}^{N} \frac{(m+r-1)!}{(m-1)!} B_{l-1, m+r+1}(\lambda), \tag{22}
\end{equation*}
$$

for any $N \geq k-(m+l)$.

## 3 Construction of the approximations

In the approximations that follow we will need the so-called signless Stirling numbers of the first kind.

### 3.1 Signless Stirling numbers of the first kind

Denote the signless Stirling numbers of the first kind as $s(n, j)$, see Pólya et al. (1983) and Comtet (1974). These numbers satisfy the recurrence

$$
s(n, j)=(n-1) s(n-1, j)+s(n-1, j-1), \quad n, j \geq 1,
$$

with the initial conditions $s(n, j)=0$ if $n \leq 0$ or $j \leq 0$, except $s(0,0)=1$. Values for these numbers can be obtained from numerous combinatorial books, for example Pólya et al. (1983).

We shall need the following properties of these Stirling numbers, $s(n, j)$. Let $H_{n}=1+\frac{1}{2}+\ldots+\frac{1}{n}$ be the harmonic numbers, then (see Comtet (1974), p. 217) $s(n, j)=0$ for $n<j$ and for all $n \geq j-1$

$$
\begin{equation*}
s(n+1,1)=n! \tag{23}
\end{equation*}
$$

$$
\begin{align*}
& s(n+1,2)=n!\left(1+\frac{1}{2}+\ldots+\frac{1}{n}\right)=n!H_{n}  \tag{24}\\
& s(n+1,3)=\frac{n!}{2}\left[H_{n}^{2}-\left(1+\frac{1}{2^{2}}+\ldots+\frac{1}{n^{2}}\right)\right] \tag{25}
\end{align*}
$$

Additionally, for $n$ fixed and variable $k, s(n+1, k+1)$ are the elementary symmetric functions of the first $n$ integers; that is, for $p=1,2, \ldots n$ we have

$$
\begin{equation*}
s(n+1, n+1-p)=\sum_{1 \leq i_{1}<i_{2}<\ldots<i_{p} \leq n} i_{1} i_{2} \ldots i_{p} . \tag{26}
\end{equation*}
$$

### 3.2 Approximating the first negative moment

Through using (14) and (15) we have

$$
\begin{equation*}
B_{1,1}(\lambda)=\frac{1}{\lambda}+\frac{1}{\lambda^{2}}+\frac{2}{\lambda^{3}}+\ldots=\sum_{r=0}^{k-1} \frac{r!}{\lambda^{r+1}}+k!B_{1, k+1}(\lambda) . \tag{27}
\end{equation*}
$$

To construct $k^{t h}$ order approximations for $\mu_{-1}$, we keep the first $k$ terms in Equation (27), ignoring $B_{1, k+1}(\lambda)$, respectively. In this way we obtain

$$
\begin{equation*}
\mu_{-1} \cong B_{1,1}(\lambda) \cong \mu_{-1}^{(k)}=\sum_{r=0}^{k-1} \frac{r!}{\lambda^{r+1}}=\frac{1}{\lambda}+\frac{1}{\lambda^{2}}+\frac{2}{\lambda^{3}}+\ldots+\frac{(k-1)!}{\lambda^{k}} \tag{28}
\end{equation*}
$$

Through using (23) it is evident that we can write $\mu_{-1}^{(k)}$ in the form of (3).
Table I: Comparison of the relative errors of the approximations for the first negative moment, against different values of $\lambda$.

|  | Relative Error |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  |  | $\mu_{-1}^{(5)}$ | $\mu_{-1}^{(\lceil 0.5 \lambda\rceil)}$ | $\mu_{-1}^{(\lceil\lambda\rceil)}$ | $T_{0}^{(1)}$ | $T_{3}^{(1)}$ | $T_{6}^{(1)}$ |
| 10 | $1.60510^{-3}$ | $1.60510^{-3}$ | $-1.21710^{-3}$ | $1.68610^{-2}$ | $1.61110^{-2}$ | $1.54610^{-2}$ |  |  |  |  |  |  |  |
| 15 | $2.60610^{-4}$ | $5.53910^{-5}$ | $-1.45910^{-5}$ | $6.11710^{-3}$ | $5.93410^{-3}$ | $5.79910^{-3}$ |  |  |  |  |  |  |  |
| 25 | $1.58010^{-5}$ | $1.71810^{-8}$ | $-1.30510^{-9}$ | $1.91210^{-3}$ | $1.85510^{-3}$ | $1.83710^{-3}$ |  |  |  |  |  |  |  |
| 50 | $4.28910^{-7}$ | $1.12410^{-17}$ | $-4.33310^{-20}$ | $4.35010^{-4}$ | $4.27510^{-4}$ | $4.26410^{-4}$ |  |  |  |  |  |  |  |
| 100 | $1.26510^{-8}$ | $6.29210^{-36}$ | $-1.93510^{-41}$ | $1.04210^{-4}$ | $1.03210^{-4}$ | $1.03110^{-4}$ |  |  |  |  |  |  |  |
| 200 | $3.84710^{-10}$ | $1.49610^{-72}$ | $-1.63310^{-79}$ | $2.55110^{-5}$ | $2.53910^{-5}$ | $2.53810^{-5}$ |  |  |  |  |  |  |  |

The accuracy of the approximations used are shown in Table I, which gives the relative error against different values of $\lambda$; all of our approximations compare favorably to the more
complex Tiku approximations $T_{j}^{(1)}$ defined in Equation (4). Already the simple approximation $\mu_{-1}^{(5)}=\frac{1}{\lambda}+\frac{1}{\lambda^{2}}+\frac{2}{\lambda^{3}}+\frac{6}{\lambda^{4}}+\frac{24}{\lambda^{5}}$, provides very good precision. The approximations $\mu_{-1}^{(3)}$ and $\mu_{-1}^{(4)}$ are also accurate. Table I shows that the approximation $\mu^{([0.5 \lambda])}$ is very accurate for $\lambda \geq 10$, and always underestimates $\mu_{-1}$. On the other hand, $\mu_{-1}^{(k)}$ with $k=\lceil\lambda\rceil$ always overestimates $\mu_{-1}$, and it is even more precise than $\mu_{-1}^{(k)}$ with $k=\lceil\lambda / 2\rceil$. This yields that any approximation $\mu_{-1}^{(k)}$, with $\left\lceil\frac{\lambda}{2}\right\rceil \leq k \leq\lceil\lambda\rceil$ provides an extremely accurate approximation.

### 3.3 Approximating the second negative moment

Using Equation (22) with $l=2$ and $m=1$, we obtain

$$
\begin{equation*}
\mu_{-2} \cong \sum_{r=0}^{N} r!B_{1, r+2}(\lambda) \tag{29}
\end{equation*}
$$

We will be constructing approximations of the form

$$
\mu_{-2} \cong \mu_{-2}^{(k)}=\sum_{r=1}^{k} \frac{a_{2,1}(r)}{\lambda^{r}},
$$

which we shall obtain from (29) with $N$ large enough (any $N>k$ would suffice), by the repeated application of the approximations in (22), and keeping all the terms that contribute to $a_{2,1}(r)$ 's with $r \leq k-1$.
Table II: Comparison of the relative errors for the approximations of the second negative moment, against different values of $\lambda$.

|  | Relative Error |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| $\lambda$ | $\mu_{-2}^{(6)}$ | $\mu_{-2}^{(\lceil 0.5 \lambda]+1)}$ | $\mu_{-2}^{(\lceil\lambda]+1)}$ | $T^{(2)}$ |
| 10 | $2.92510^{-2}$ | $2.92510^{-2}$ | $-1.15810^{-2}$ | $9.35510^{-2}$ |
| 15 | $4.14810^{-3}$ | $1.42510^{-3}$ | $-2.58710^{-4}$ | $2.99310^{-1}$ |
| 25 | $2.31610^{-4}$ | $7.84010^{-7}$ | $-4.67110^{-8}$ | $8.33810^{-3}$ |
| 50 | $6.25010^{-6}$ | $1.15710^{-15}$ | $-3.82410^{-18}$ | $1.80810^{-3}$ |
| 100 | $1.84910^{-7}$ | $1.46010^{-33}$ | $-4.07710^{-39}$ | $4.24410^{-4}$ |
| 200 | $5.63910^{-9}$ | $7.87110^{-70}$ | $-8.02310^{-82}$ | $1.03010^{-4}$ |

Rewrite (22) as

$$
\begin{equation*}
B_{l, m}(\lambda) \cong \sum_{r \geq 0} \frac{(m+r-1)!}{(m-1)!} B_{l-1, m+r+1}(\lambda) \tag{30}
\end{equation*}
$$

and, respectively, (29) as

$$
\begin{equation*}
\mu_{-2} \cong \sum_{r \geq 0} r!B_{1, r+2}(\lambda) \tag{31}
\end{equation*}
$$

Hence, from (31) and (30) we obtain

$$
\mu_{-2} \cong \sum_{r \geq 0} r!B_{1, r+2}(\lambda) \cong \sum_{r \geq 0} r!\sum_{s \geq 0} \frac{1}{\lambda^{s+r+2}} \frac{(s+r+1)!}{(r+1)!}=\sum_{r \geq 0} \frac{1}{r+1} \sum_{s \geq 0} \frac{1}{\lambda^{s+r+2}}(s+r+1)!.
$$

Let $t=r+s+1$; using Equation (24) we obtain

$$
\begin{equation*}
\mu_{-2} \cong \sum_{r \geq 0} \sum_{t \geq r+1} \frac{t!}{(r+1) \lambda^{t+1}}=\sum_{t \geq 1} \frac{t!}{\lambda^{t+1}} \sum_{r=0}^{t-1} \frac{1}{r+1}=\sum_{t \geq 1} \frac{t!H_{t}}{\lambda^{t+1}}=\sum_{t \geq 2} \frac{s(t, 2)}{\lambda^{t}} . \tag{32}
\end{equation*}
$$

The $k^{\text {th }}$ order approximation to $\mu_{-2}$ is obtained when we keep the first $k$ terms in the right-hand side of (32).

In Table II we discover that $\mu_{-2}^{([0.5 \lambda]+1)}, \mu_{-2}^{([\lambda]+1)}$ and even the very simple approximation $\mu_{-2}^{(6)}$ compare very favorable against Tiku's approximation $T^{(2)}$. We find that $\mu_{-2}^{[\lambda]+1}$ overestimates $\mu_{-2}$, whereas $\mu_{-2}^{(\lceil 0.5 \lambda]+1)}$ understimates, this suggests any approximation $\mu_{-2}^{(k)}$ with $\left\lceil\frac{\lambda}{2}\right\rceil+1 \leq k \leq\lceil\lambda\rceil+1$ would provide an extremely accurate approximation.

### 3.4 Approximating the third negative moment

To approximate the third negative moment, $\mu_{-3}$, we express $B_{3,1}(\lambda)$ in a similar way:

$$
\begin{aligned}
\mu_{-3} & \cong B_{3,1}(\lambda) \cong \sum_{r \geq 0} r!B_{2, r+2}(\lambda) \cong \sum_{r \geq 0} r!\sum_{s \geq 0} B_{1, r+s+3}(\lambda) \frac{(r+s+1)!}{(r+1)!} \\
& \cong \sum_{r \geq 0} \frac{1}{r+1} \sum_{s \geq 0}(r+s+1)!\sum_{t \geq 0} \frac{1}{\lambda^{t+r+s+3}} \frac{(r+s+t+2)!}{(r+s+2)!} \\
& =\sum_{r \geq 0} \frac{1}{r+1} \sum_{s \geq 0} \frac{1}{r+s+2} \sum_{t \geq 0} \frac{1}{\lambda^{t+r+s+3}}(r+s+t+2)!.
\end{aligned}
$$

Let $u=r+s+t+2$ and $s^{\prime}=r+s+1$, so that

$$
\begin{equation*}
\mu_{-3} \cong \sum_{u \geq 2} \frac{u!}{\lambda^{u+1}} \sum_{r=0}^{u-2} \sum_{s^{\prime}=r+1}^{u-1} \frac{1}{(r+1)\left(s^{\prime}+1\right)} \tag{33}
\end{equation*}
$$

Since

$$
\begin{align*}
H_{u}^{2} & =\sum_{r=0}^{u-1} \frac{1}{r+1} \sum_{s=0}^{u-1} \frac{1}{s+1}=\sum_{r=0}^{u-2} \frac{1}{r+1} \sum_{s=r+1}^{u-1} \frac{1}{s+1}+\sum_{r=0}^{u-1}\left[\frac{1}{r+1}\right]^{2}+\sum_{r=1}^{u-1} \frac{1}{r+1} \sum_{s=0}^{r-1} \frac{1}{s+1} \\
& =2 \sum_{r=0}^{u-2} \frac{1}{r+1} \sum_{s=r+1}^{u-1} \frac{1}{s+1}+\left(1+\frac{1}{2^{2}}+\frac{1}{3^{2}}+\ldots+\frac{1}{u^{2}}\right) \tag{34}
\end{align*}
$$

we have from (33) and (25)

$$
\begin{equation*}
\mu_{-3} \cong \sum_{u \geq 2} \frac{u!}{2 \lambda^{u+1}}\left[H_{u}^{2}-\left(1+\frac{1}{2^{2}}+\ldots+\frac{1}{u^{2}}\right)\right]=\sum_{u \geq 2} \frac{s(u+1,3)}{\lambda^{u+1}}=\sum_{u \geq 3} \frac{s(u, 3)}{\lambda^{u}} . \tag{35}
\end{equation*}
$$

The $k^{t h}$ order approximation $\mu_{-3}^{(k)}$ is obtained by keeping the first $k$ terms in (35).
Table III: Comparison of the relative errors for approximations of the third negative moment, against different values of $\lambda$.

|  | Relative Error |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\lambda$ | $\mu_{-3}^{(7)}$ | $\mu_{-3}^{([0.5 \lambda]+2)}$ | $\mu_{-3}^{(\Gamma \lambda\rceil+2)}$ | $T^{(3)}$ |  |
| 10 | $1.94110^{-1}$ | $1.94110^{-1}$ | $-2.83410^{-2}$ | $3.15910^{-1}$ |  |
| 15 | $3.33710^{-2}$ | $1.72010^{-2}$ | $-1.44010^{-3}$ | $9.70410^{-2}$ |  |
| 25 | $1.66910^{-3}$ | $1.71910^{-5}$ | $-5.92310^{-7}$ | $2.28710^{-2}$ |  |
| 50 | $4.53410^{-5}$ | $5.56710^{-14}$ | $-1.36010^{-16}$ | $4.70010^{-3}$ |  |
| 100 | $1.35810^{-6}$ | $1.59010^{-31}$ | $-3.85510^{-37}$ | $1.08110^{-3}$ |  |
| 200 | $4.16910^{-8}$ | $1.96310^{-67}$ | $-1.92410^{-79}$ | $2.59710^{-4}$ |  |

In Table III we discover that all approximations compare favorably against Tiku's approximation $T^{(3)}$, and if we had and approximation of the form $\mu_{-3}^{(k)}$ with $\lceil 0.5 \lambda\rceil+2 \leq k \leq$ $\lceil\lambda\rceil+2$, the approximation would be extremely accurate.

### 3.5 General Case

For the $l^{\text {th }}$ order negative moment we approximate $B_{l, 1}(\lambda)$. This can be obtained by applying the same methodology:
$\mu_{-l} \cong \tilde{\mu}_{-l}=\sum_{r_{1} \geq 0} \frac{1}{r_{1}+1} \sum_{r_{2} \geq 0} \frac{1}{r_{1}+r_{2}+2} \ldots \sum_{r_{l-1} \geq 0} \frac{1}{r_{1}+r_{2}+\ldots+r_{l-1}+l-1} \sum_{r_{l} \geq 0} \frac{\left(r_{1}+r_{2}+\ldots+r_{l}+l-1\right)!}{\lambda^{r_{1}+r_{2}+\ldots+r_{l}+l}}$.

Let $r_{m}^{\prime}=\sum_{i=1}^{m} r_{i}+(m-1)$, for $2 \leq r \leq l$.
Thus, we have

$$
\tilde{\mu}_{-l}=\sum_{r_{l}^{\prime} \geq l-1} \frac{r_{l}^{\prime}!}{\lambda_{l}^{r_{l}^{\prime}+1}} \sum_{r_{1}=0}^{r_{l}^{\prime}-(l-1)} \sum_{r_{2}^{\prime}=r_{1}+1}^{r_{l}^{\prime-}(l-2)} \sum_{r_{3}^{\prime}=r_{2}^{\prime}+1}^{r_{l}^{\prime}-(l-3)} \cdots \sum_{r_{l-1}^{\prime}=r_{l-2}^{\prime}+1}^{r_{l}^{\prime}-1} \frac{1}{\left(r_{1}+1\right)\left(r_{2}^{\prime}+1\right) \ldots\left(r_{l-1}^{\prime}+1\right)} .
$$

Let $t_{1}=r_{1}+1$ and $t_{m}=r_{m}^{\prime}+1,2 \leq m \leq l$. Hence, using Equation (26)

$$
\begin{align*}
\tilde{\mu}_{-l} & =\sum_{t_{l} \geq l-1} \frac{t_{l}!}{\lambda^{t_{l}+1}} \sum_{t_{1}=1}^{t_{l}-(l-2)} \sum_{t_{2}=t_{1}}^{t_{l}-(l-1)} \sum_{t_{3}=t_{2}}^{t_{l}-(l-2)} \cdots \sum_{t_{l-1}=t_{l-2}}^{t_{l}-2} \frac{1}{t_{1} t_{2} \ldots t_{l-1}} \\
& =\sum_{t_{l} \geq l-1} \frac{s\left(t_{l}+1, l\right)}{\lambda^{t_{l}+1}}=\sum_{r \geq l} \frac{s(r, l)}{\lambda^{r}} . \tag{36}
\end{align*}
$$

Similarly to the above, the $k^{t h}$ order approximation to $\mu_{-l}$ is obtained when we keep the first $k$ terms in (36).

### 3.6 Approximating non-integer negative moments

Assume that $\alpha>0$ is not an integer. Set $l=\lfloor\alpha\rfloor$ and let $\theta=\alpha-l$ with $0 \leq \theta \leq 1$. Accurate approximations for non-integer negative moments can be obtained through using the following approximation:

$$
\begin{equation*}
\mu_{-\alpha}^{(t+\alpha)}=\sum_{r=0}^{t} \frac{s(r+1, l)+\theta[s(r+l+1, l+1)-s(r+1, l)]}{\lambda^{\alpha+r}} . \tag{37}
\end{equation*}
$$

Table IV clearly demonstrates that all of the approximations for the non-integer negative moments provide very accurate approximations for $\lambda \geq 50$, with a relative error less than 0.003. However, it can also be seen that for larger values of $\alpha$ (e.g. $\alpha=5.1$ and 5.5) the approximations are nor very accurate for small $\lambda$. This can be explained by the fact that the approximations for the larger integer negative moments are only accurate for fairly large values of $\lambda$. Additionally, we find that the approximations are less accurate as $\theta$ departs further away from zero or one.

Table IV: Comparison of relative errors of the non-integer negative moments for different values of $\alpha$ versus different values of $\lambda$, for $t=5$.

| $\lambda$ | Relative Error |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $\mu_{-0.1}$ | $\mu_{-0.5}$ | $\mu_{-2.1}$ | $\mu_{-2.5}$ | $\mu_{-5.1}$ | $\mu_{-5.5}$ |
| 10 | -6.872 $10^{-3}$ | -1.959 $10^{-2}$ | $8.46310^{-4}$ | $2.12210^{-3}$ | $8.30310^{-1}$ | $8.91610^{-1}$ |
| 15 | $-3.85210^{-3}$ | -1.095 $10^{-2}$ | -5.835 $10^{-3}$ | -1.633 $10^{-2}$ | $5.83210^{-1}$ | $7.44110^{-1}$ |
| 25 | -2.072 $10^{-3}$ | -5.834 $10^{-2}$ | -3.299 10-3 | -9.216 $10^{-3}$ | $1.28810^{-2}$ | $1.76210^{-1}$ |
| 50 | -9.629 10-3 | -2.692 $10^{-2}$ | -1.225 $10^{-3}$ | -3.416 $10^{-3}$ | -1.853 $10^{-3}$ | -5.351 $10^{-3}$ |
| 100 | $-4.65210^{-3}$ | -1.296 $10^{-2}$ | -5.245 $10^{-4}$ | -1.461 $10^{-3}$ | -7.002 10-4 | -1.946 $10^{-3}$ |
| 200 | $-2.28710^{-4}$ | -6.363 $10^{-3}$ | -2.429 $10^{-4}$ | -6.756 $10^{-4}$ | -2.835 $10^{-4}$ | -7.878 10-4 |

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