# A combinatorial and probabilistic study of initial and end heights of descents in samples of geometrically distributed random variables and in permutations 

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In words, generated by independent geometrically distributed random variables, we study the $l$ th descent, which is, roughly speaking, the $l$ th occurrence of a neighbouring pair $a b$ with $a>b$. The value $a$ is called the initial height, and $b$ the end height. We study these two random variables (and some similar ones) by combinatorial and probabilistic tools.

## 1 Introduction

Let $X$ be a random variable (RV), distributed according to the geometric distribution with parameter $p(\operatorname{geom}(p))$ : $\mathbb{P}(X=k)=p q^{k-1}$, with $q=1-p$. We consider a sequence $X_{1} X_{2} \ldots X_{n}$ of independent RVs. We also speak about words $a_{1} \ldots a_{n}$; there is some interest of combinatorial parameters of such words, generated by independent geometric random variables. In this paper we continue the study of descents.

In a word $w_{1} a b w_{2}$ we say that $a b$ is the $l$ th descent, if $a>b$ (strict model) or $a \geq b$ (weak model), and the initial word $w_{1} a$ has $l-1$ descents. Furthermore, we refer to $a$ as the initial height and to $b$ as the end height. Equivalently, we use the notions initial value and end value.

In (5), these random variables where studied, but only for $l=1$, i.e., the first descent. Here, we are able to deal with the general case.

This paper uses a generating functions approach and a probabilistic approach. The results complement each other, but are not disjoint. Of course, both are very useful and interesting.

The generating function approach is as follows: First, we construct a generating function $F(z, v, u)$ in 3 variables, $z, v, u$, where $z$ marks the length of the word, $v$ the number of descents, and $u$ the last letter of the word. In other words,

$$
F(z, v, u)=\sum_{n, j \geq 1, l \geq 0} \mathbb{P}[\text { a word of length } n \text { has } l \text { descents and last letter } j] .
$$

[^0]Occasionally, it is clearer to write $F(u)$ only.
Once this is achieved, we construct a new generating function ( $G$ for initial height, $H$ for end height), by attaching a descent (which is a simple substitution, since the variable $u$ "remembers" the last letter) and an arbitrary rest. In this way, we have a generating function, where the variable $u$ no longer codes the last letter, but the initial height (resp. end height) of the $l$ th descent.
The quantities that we get for $l=1$ coincide (of course!) with the older paper; however, they come out in different forms. To show formally that they are the same, one uses identities from $q$-analysis, such as Heine's transformation formula. This was demonstrated extensively in (5).

We also consider the analogous questions for ascents; the motivation is that, in (4), the last descent was studied. In the reversed word, the last descent becomes the first ascent, and now we have developed the machinery to deal in general with the $l$ th ascent.

The probabilistic approach works as follows. We start from an infinite sequence of geometric random variables, with parameter $p(\operatorname{geom}(p))$. First, we consider the successive descents as a Markov chain, related to initial and end values of each descent. Next, we use this Markov chain to obtain the distribution of initial and end values of first and second descents. Then, the first moments of the first and second descents initial values are analyzed by intensive use of some combinatorial identities. Next, we obtain the asymptotic distribution of the number of descents initial values in some interval. Finally, starting from $n$ geom $(p)$ RV, as $q \rightarrow 1$, we can derive the asymptotic properties of first and second descents, in a large permutation.

In this part, only the strict model and the descents will be considered.
The explicit forms of the distribution of the descents become very complicated when going from first to second etc. descent. However, a stationary distribution, which is very simple, is rapidly approached. It is given by

$$
\frac{1+q}{q} p q^{i-1}\left(1-q^{i-1}\right)
$$

for the initial height, and by

$$
\frac{1+q}{q} p q^{2 i-1}
$$

for the end height. There is an intuitive explanation of them: The first one is the conditional probability that we have a pair $i j$, given that it is a descent, and the second one that we have a pair $j i$, again given that it is a descent.

Several useful combinatorial identities, derived from Heine's formula, are given in the Appendix.
We will need notation from $q$-analysis; the most important ones are $(x)_{n}:=(1-x)(1-q x) \ldots\left(1-x q^{n-1}\right)$, and $\left[\begin{array}{l}n \\ k\end{array}\right]=\frac{(q)_{n}}{(q)_{k}(q)_{n-k}}$ (Gaussian coefficients). The relevant formulæ can be found in (1).

## Combinatorial Analysis

## 2 Descents: the weak model

Let $f_{i}(u)$ be the generating function with $\left[z^{n} u^{j}\right] f_{i}(u)$ is the probability that a word of length $n \geq 1$ has $i$ descents, and that the last letter is $j$. (Only the dependency on the variable $u$ has been made explicit.)

Here is the recursion for $i \geq 1$.

$$
f_{i}(u)=\frac{p u z}{1-q u} f_{i-1}(1)-\frac{p u z}{1-q u} f_{i-1}(u q)+\frac{p u z}{1-q u} f_{i}(u q)
$$

Now let

$$
F(u)=F(z, v, u)=\sum_{i \geq 0} f_{i}(u) v^{i}
$$

Then we get, by summing up,

$$
F(u)-f_{0}(u)=\frac{p u v z}{1-q u} F(1)-\frac{p u v z}{1-q u} F(u q)+\frac{p u z}{1-q u} F(u q)-\frac{p u z}{1-q u} f_{0}(u q) .
$$

But

$$
f_{0}(u)=\frac{p u z}{1-q u}+\frac{p u z}{1-q u} f_{0}(u q)
$$

and so

$$
F(u)=\frac{p u z}{1-q u}+\frac{p u v z}{1-q u} F(1)+\frac{p u z(1-v)}{1-q u} F(u q) .
$$

This functional equation can be solved by iterating it:

$$
\begin{equation*}
F(z, v, u)=\frac{\sum_{k \geq 1} \frac{(p u z)^{k}(1-v)^{k} q\binom{k}{2}}{(q u)_{k}}}{1-v \sum_{k \geq 0} \frac{(p z)^{k}(1-v)^{k} q\binom{k}{2}}{(q)_{k}}} \tag{2.1}
\end{equation*}
$$

Now we turn to the end heights. As indicated already, we make one down step, record its height with the $u$-variable, and then attach anything $(1 /(1-z))$. This gives the generating function

$$
H(z, v, u)=\frac{p u v z}{1-q u}[F(z, v, 1)-F(z, v, u q)] \frac{1}{1-z}+\frac{p z u}{1-q u}
$$

The next step is to look at the behaviour for $n \rightarrow \infty$. Intuitively, it is quite clear, there should be a limit, since what happens at the $l$ th descent must become independent of letters very far to the right. Indeed, we see that there is a simple pole at $z=1$, and the generating function

$$
\Psi(v, u)=\frac{p u v}{1-q u}[F(1, v, 1)-F(1, v, u q)]
$$

that is obtained from $H(z, v, u)$ by dropping the factor $1 /(1-z)$, and the irrelevant additive term and replacing $z=1$, contains all the information. The coefficient of $v^{l}$ is a probability generating function in the variable $u$ alone. Indeed, it is easy to check that $\Psi(v, 1)=\frac{v}{1-v}$. Explicit expressions become very messy, but at least the instance $l=1$ (which was studied in (5)) is manageable:

$$
\begin{aligned}
{\left[v^{1}\right] \Psi(v, u) } & =\frac{p u}{1-q u}[F(1,0,1)-F(1,0, u q)] \\
& =\frac{p u}{1-q u}\left[\sum_{k \geq 1} \frac{p^{k} q^{\binom{k}{2}}}{(q)_{k}}-\sum_{k \geq 1} \frac{(p q u)^{k} q^{\binom{k}{2}}}{\left(q^{2} u\right)_{k}}\right] .
\end{aligned}
$$

All moments can be derived from this by differentiations. Let us just do this for the average (first moment):

$$
\begin{align*}
\mathbb{E}[\text { first descent }] & \sim \frac{1}{p}-\left.\frac{\partial}{\partial u} \sum_{k \geq 1} \frac{(p q u)^{k} q^{\binom{k}{2}}}{\left(q^{2} u\right)_{k}}\right|_{u=1} \\
& \sim \frac{1}{p}-\sum_{k \geq 1} \frac{k(p q)^{k} q^{\binom{k}{2}}}{\left(q^{2}\right)_{k}}+\sum_{k \geq 1} \frac{(p q)^{k} q^{\binom{k}{2}}}{\left(q^{2}\right)_{k}} \sum_{i=2}^{k+1} \frac{q^{i}}{1-q^{i}} \\
& \sim \frac{1}{p}-\sum_{k \geq 2} \frac{(k-1) p^{k} q^{\binom{k}{2}}}{(q)_{k}}-\sum_{k \geq 2} \frac{p^{k} q^{\binom{k}{2}}}{(q)_{k}} \sum_{i=2}^{k} \frac{q^{i}}{1-q^{i}} \tag{2.2}
\end{align*}
$$

Note that this average was given as

$$
\begin{equation*}
\frac{1}{p}-\sum_{h \geq 0}(h+1) p q^{2 h+1}(-p)_{h} \tag{2.3}
\end{equation*}
$$

in (5).
For the limiting behaviour of this, as $q \rightarrow 1$, we should consider, in order to get a meaningful result, $\lim _{q \rightarrow 1}(1-q) \mathbb{E}$ [first descent], which evaluates to

$$
1-\sum_{k \geq 1} \frac{1}{k!} \sum_{i=2}^{k} \frac{1}{i}=1-\sum_{k \geq 1} \frac{H_{k}}{k!}+\sum_{k \geq 1} \frac{1}{k!}=e-\sum_{k \geq 1} \frac{H_{k}}{k!} .
$$

A similar compution can be done for general $l$, and the resulting generating function is

$$
v-\frac{1}{1-v}+\frac{1}{1-v e^{1-v}}-\frac{v}{1-v e^{1-v}} \sum_{k \geq 1} \frac{H_{k}(1-v)^{k-1}}{k!}
$$

(the coefficient of $v^{1}$ is the previous expression).
Now we turn to the initial heights. The same approach applies, but the substitution is even simpler. We get

$$
G(z, u, v)=v z[F(z, v, u)-F(z, v, u q)] \frac{1}{1-z}+\frac{p z u}{1-q u}
$$

The limit for $n \rightarrow \infty$ leads then to the generating function

$$
\Phi(v, u)=v[F(1, v, u)-F(1, v, u q)] .
$$

Let us compute again the instance $l=1$; the coefficient of $v^{1}$ is particularly simple:

$$
\left[v^{1}\right] \Phi(v, u)=F(1,0, u)-F(1,0, u q)
$$

From this, we find that the average is asymptotic to

$$
\begin{equation*}
\left.\frac{\partial}{\partial u}\left[\sum_{k \geq 1} \frac{(p u)^{k} q^{\binom{k}{2}}}{(q u)_{k}}-\sum_{k \geq 1} \frac{(p q u)^{k} q^{\binom{k}{2}}}{\left(q^{2} u\right)_{k}}\right]\right|_{u=1} \tag{2.4}
\end{equation*}
$$

This checks with the expression

$$
\begin{equation*}
\sum_{h \geq 0}(h+1) p q^{h+1}(-p)_{h}-\sum_{h \geq 0}(2 h+1) p q^{h+1}(-p)_{h} \tag{2.5}
\end{equation*}
$$

given in (5).
For the limit $\lim _{q \rightarrow 1}(1-q) \mathbb{E}[$ first descent $]$ we get the generating function

$$
\frac{1}{v\left(1-v e^{1-v}\right)}-\frac{1}{v(1-v)}-\frac{v}{1-v e^{1-v}} \sum_{k \geq 1} \frac{H_{k}(1-v)^{k-1}}{k!}
$$

The coefficient of $v$ in this is

$$
e-1-\sum_{k \geq 1} \frac{H_{k}}{k!}
$$

## 3 Descents: the strict model

Computations are similar; we only give the key steps. The functional equation is

$$
F(u)=\frac{p u z}{1-q u}+\frac{p u v z}{1-q u} F(1)+\frac{p z(1-v)}{q(1-q u)} F(u q) .
$$

Thus

$$
\begin{equation*}
F(z, v, u)=u \sum_{k \geq 1} \frac{(p z)^{k}(1-v)^{k-1}}{(u q)_{k}}(1+v F(z, v, 1))=\frac{u \sum_{k \geq 1} \frac{(p z)^{k}(1-v)^{k}}{(u q)_{k}}}{1-v \sum_{k \geq 0} \frac{(p z)^{k}(1-v)^{k}}{(q)_{k}}} \tag{3.1}
\end{equation*}
$$

Also,

$$
H(z, v, u)=\left[\frac{p u v z}{1-q u} F(z, v, 1)-\frac{p v z}{q(1-q u)} F(z, v, u q)\right] \frac{1}{1-z}+\frac{p z u}{1-q u}
$$

Hence

$$
\Psi(v, u)=\frac{p u v}{1-q u} F(1, v, 1)-\frac{p v}{q(1-q u)} F(1, v, u q) .
$$

And

$$
\begin{aligned}
{\left[v^{1}\right] \Psi(v, u) } & =\frac{p u}{1-q u} F(1,0,1)-\frac{p}{q(1-q u)} F(1,0, u q) \\
& =\frac{p u}{1-q u} \sum_{k \geq 1} \frac{p^{k}}{(q)_{k}}-\frac{p u}{1-q u} \sum_{k \geq 1} \frac{p^{k}}{\left(u q^{2}\right)_{k}}
\end{aligned}
$$

The expectation can be obtained by differentiation, followed by $u=1$ :

$$
\begin{equation*}
\frac{1}{p}-\sum_{k \geq 1} \frac{p^{k}}{(q)_{k}} \sum_{i=2}^{k} \frac{q^{i}}{1-q^{i}} \tag{3.2}
\end{equation*}
$$

The formula given in (5) is

$$
\begin{equation*}
\frac{1}{p}-\sum_{h \geq 1} \frac{h q^{2 h-1}}{(p q)_{h}} \tag{3.3}
\end{equation*}
$$

Now, let us compute the limit of $(1-q) \mathbb{E}$ :

$$
1-\sum_{k \geq 1} \frac{H_{k}-1}{k!}=e-\sum_{k \geq 1} \frac{H_{k}}{k!}
$$

And in general:

$$
\frac{1}{1-v e^{1-v}}-\frac{1}{1-v}-\frac{v}{1-v e^{1-v}} \sum_{k \geq 2} \frac{H_{k}(1-v)^{k-1}}{k!}
$$

Now, for the initial heights, we must consider

$$
G(z, v, u)=v z\left[F(z, v, u)-\frac{1}{q} F(z, v, q u)\right] \frac{1}{1-z}+\frac{p z}{1-q u}
$$

and

$$
\Phi(v, u)=v\left[F(1, v, u)-\frac{1}{q} F(1, v, q u)\right] .
$$

Furthermore, to look at the first decent,

$$
\left[v^{1}\right] \Phi(v, u)=F(1,0, u)-\frac{1}{q} F(1,0, q u)=u \sum_{k \geq 1} \frac{p^{k}}{(q u)_{k}}-u \sum_{k \geq 1} \frac{p^{k}}{\left(q^{2} u\right)_{k}}
$$

From this, the average, obtained by differentiation, is

$$
\begin{align*}
1+ & \sum_{k \geq 1} \frac{p^{k}}{(q)_{k}} \sum_{i=1}^{k} \frac{q^{i}}{1-q^{i}}-\sum_{k \geq 1} \frac{p^{k}}{\left(q^{2}\right)_{k}} \sum_{i=2}^{k+1} \frac{q^{i}}{1-q^{i}} \\
& =1+\sum_{k \geq 1} \frac{p^{k}}{(q)_{k}} \sum_{i=1}^{k} \frac{q^{i}}{1-q^{i}}-\sum_{k \geq 2} \frac{p^{k}}{(q)_{k}} \sum_{i=2}^{k} \frac{q^{i}}{1-q^{i}} \\
& =1+\frac{q}{p} \sum_{k \geq 1} \frac{p^{k}}{(q)_{k}} \\
& =1+\frac{q}{p}\left(-1+\frac{1}{(p)_{\infty}}\right) . \tag{3.4}
\end{align*}
$$

The last simplification was by (A.6). The version given in (5) is

$$
\begin{equation*}
1+\sum_{h \geq 1} \frac{h p q^{h-1}}{(p q)_{h}}-\sum_{h \geq 1} \frac{h p q^{2 h-1}}{(p q)_{h}} \tag{3.5}
\end{equation*}
$$

The limiting function $\lim _{q \rightarrow 1}(1-q) \mathbb{E}$ is

$$
\frac{v}{1-v e^{1-v}}\left[\sum_{k \geq 1} \frac{(1-v)^{k} H_{k}}{k!}-\sum_{k \geq 1} \frac{(1-v)^{k}\left(H_{k+1}-1\right)}{(k+1)!}\right]
$$

## 4 Ascents: the strict model

First, we consider the case where only $a<b$ is an ascent.
Again, the treatment is very similar to before, so we only give the key steps.
Let $f_{i}(u)$ be the generating function with $\left[z^{n} u^{j}\right] f_{i}(u)$ is the probability that a word of length $n \geq 1$ has $i$ ascents, and that the last letter is $j$.

Here is the recursion for $i \geq 1$ :

$$
f_{i}(u)=\frac{p u z}{1-q u} f_{i}(1)-\frac{p u z}{1-q u} f_{i}(u q)+\frac{p u z}{1-q u} f_{i-1}(u q)
$$

Now let

$$
F(u)=F(z, v, u)=\sum_{i \geq 0} f_{i}(u) v^{i}
$$

Then we get

$$
F(u)=\frac{p u z}{1-q u}+\frac{p u z}{1-q u} F(1)+\frac{p u z(v-1)}{1-q u} F(u q) .
$$

This functional equation can be solved by iterating it:

$$
\begin{equation*}
F(z, v, u)=\frac{\sum_{k \geq 1} \frac{(p u z)^{k}(v-1)^{k-1} q\binom{k}{2}}{(q u)_{k}}}{1-\sum_{k \geq 1} \frac{(p z)^{k}(v-1)^{k-1} q^{k}\binom{k}{2}}{(q)_{k}}} . \tag{4.1}
\end{equation*}
$$

Further,

$$
H(z, v, u)=\frac{p u z v}{1-q u} F(z, v, u q) \frac{1}{1-z}+\frac{p u z}{1-q u}
$$

and

$$
\Psi(v, u)=\frac{p u v}{1-q u} F(1, v, u q)
$$

The case of the first accent, i.e., the coefficient of $v^{1}$, is thus

$$
\left[v^{1}\right] \Psi(v, u)=\frac{p u}{1-q u} F(1,0, u q)=\frac{p u}{1-q u} \frac{\sum_{k \geq 1} \frac{(p q u)^{k}(-1)^{k-1}{ }^{2}\binom{k}{2}}{\left(q^{2} u\right)_{k}}}{1-\sum_{k \geq 1} \frac{(p)^{k}(-1)^{k-1} q\binom{k}{2}}{(q)_{k}}}=\frac{\sum_{k \geq 2} \frac{(p u)^{k}(-1)^{k} q\binom{k}{2}}{(q u)_{k}}}{\sum_{k \geq 0} \frac{(p)^{k}(-1)^{k} q^{k}\binom{k}{2}}{(q)_{k}}}
$$

Notice that

$$
\sum_{k \geq 0} \frac{(p)^{k}(-1)^{k} q^{\binom{k}{2}}}{(q)_{k}}=(p)_{\infty}
$$

And therefore the average of the end height of the first ascent tends to

$$
\begin{equation*}
\frac{1}{(p)_{\infty}} \sum_{k \geq 2} \frac{k p^{k}(-1)^{k} q^{\binom{k}{2}}}{(q)_{k}}+\frac{1}{(p)_{\infty}} \sum_{k \geq 2} \frac{p^{k}(-1)^{k} q^{\binom{k}{2}}}{(q)_{k}} \sum_{i=1}^{k} \frac{q^{i}}{1-q^{i}} \tag{4.2}
\end{equation*}
$$

This quantity appears in (4) in the form

$$
\begin{equation*}
\frac{q}{p}+\frac{1}{(p)_{\infty}} \sum_{h \geq 0}(h+2) p q^{2 h+1}(p)_{h} \tag{4.3}
\end{equation*}
$$

Furthermore,

$$
G(z, v, u)=v z F(z, v, u q) \frac{1}{1-z}+\frac{p u z}{1-q u}
$$

and

$$
\Phi(v, u)=v F(1, v, u q)
$$

Also,

$$
\left[v^{1}\right] \Phi(v, u)=F(1,0, u q)=\frac{1}{(p)_{\infty}} \sum_{k \geq 1} \frac{(p q u)^{k}(-1)^{k-1} q^{\binom{k}{2}}}{\left(q^{2} u\right)_{k}}
$$

And therefore the average of the initial height of the first ascent tends to

$$
\begin{equation*}
\frac{1}{(p)_{\infty}} \sum_{k \geq 2} \frac{(k-1) p^{k}(-1)^{k} q^{\binom{k}{2}}}{(q)_{k}}+\frac{1}{(p)_{\infty}} \sum_{k \geq 2} \frac{p^{k}(-1)^{k} q^{\binom{k}{2}}}{(q)_{k}} \sum_{i=2}^{k} \frac{q^{i}}{1-q^{i}} \tag{4.4}
\end{equation*}
$$

The version given in the paper (5) is

$$
\begin{equation*}
\frac{1}{(p)_{\infty}} \sum_{h \geq 0}(h+1) p q^{2 h+1}(p)_{h} \tag{4.5}
\end{equation*}
$$

## 5 Ascents: the weak model

We only collect the relevant formulæ here:

$$
\begin{gathered}
F(u)=\frac{p u z}{1-q u}+\frac{p u z}{1-q u} F(1)+\frac{p z(v-1)}{q(1-q u)} F(u q) \\
F(z, v, u)=\frac{u \sum_{k \geq 1} \frac{(p z)^{k}(v-1)^{k-1}}{(q u)_{k}}}{1-\sum_{k \geq 1} \frac{(p z)^{k}(v-1)^{k-1}}{(q)_{k}}} \\
H(z, v, u)=\frac{p z v}{q(1-q u)} F(z, v, u q) \frac{1}{1-z}+\frac{p z}{1-q u} \\
\Psi(v, u)=\frac{p v}{q(1-q u)} F(1, v, u q) \\
G(z, v, u)=\frac{z v}{q} F(z, v, u q) \frac{1}{1-z}+\frac{p z}{1-q u} \\
\Phi(v, u)=\frac{v}{q} F(1, v, u q)
\end{gathered}
$$

## Probabilistic Analysis

We will analyze the descents in the strict model with probabilistic tools. We start from an infinite sequence of geom ( $p$ ) RVs.

## 6 Markov chains

In this section, we consider the successive descents as a Markov chain, related to initial and end values of each descent.

Let

$$
\begin{aligned}
\pi(i) & :=p q^{i-1}, \\
P(i) & :=\sum_{j \geq i} \pi(j)=q^{i-1}, \\
I_{k} & :=\text { beginning of the } k \text { th descent (initial height), } \\
J_{k} & :=\text { end of the } k \text { th descent (end height), } \\
I_{k} & >J_{k}, I_{k+1} \geq J_{k}, I_{k} \geq 2 .
\end{aligned}
$$

By convention, $J_{0}$ is the first geom $(p) \mathrm{RV}$. We have

$$
\mathbb{P}\left[I_{2}=i_{2}, J_{2}=j_{2} \mid I_{1}=i_{1}, J_{1}=j_{1}\right]=\sum_{l \geq 0} \sum_{j_{1} \leq k_{2} \leq \cdots \leq k_{l} \leq i_{2}} \pi\left(k_{2}\right) \ldots \pi\left(k_{l}\right) \pi\left(i_{2}\right) \pi\left(j_{2}\right)
$$

with the conventions

$$
\begin{aligned}
& l=0: i_{2} \equiv j_{1} \\
& l=1: i_{2} \geq j_{1}
\end{aligned}
$$

This is independent of $i_{1}$. Set

$$
A(a, b, t):=\sum_{a \leq k_{1} \leq \cdots \leq k_{t} \leq b} q^{k_{1}+\cdots+k_{t}}
$$

and

$$
B(a, b):=\sum_{t \geq 0}\left(\frac{p}{q}\right)^{t} A(a, b, t)
$$

Then

$$
\mathbb{P}\left[I_{2}=i_{2}, J_{2}=j_{2} \mid J_{1}=j_{1}\right]=\llbracket i_{2}=j_{1} \rrbracket \pi\left(j_{2}\right)+B\left(j_{1}, i_{2}\right) \pi\left(i_{2}\right) \pi\left(j_{2}\right)
$$

But we know that

$$
\sum_{0 \leq k_{1} \leq \cdots \leq k_{t} \leq n} q^{k_{1}+\cdots+k_{t}}=\frac{1}{(q)_{t}} \sum_{j=0}^{t}(-1)^{j}\left[\begin{array}{l}
t \\
j
\end{array}\right] q^{j n+\binom{j+1}{2}}=\left[\begin{array}{c}
n+t \\
t
\end{array}\right]
$$

Therefore

$$
\begin{aligned}
A(a, b, t) & :=\sum_{a \leq k_{1} \leq \cdots \leq k_{t} \leq b} q^{k_{1}+\cdots+k_{t}} \\
& =q^{a t} \sum_{0 \leq k_{1} \leq \cdots \leq k_{t} \leq b-a} q^{k_{1}+\cdots+k_{t}} \\
& =q^{a t}\left[\begin{array}{c}
b-a+t \\
b-a
\end{array}\right]
\end{aligned}
$$

We will use

$$
\sum_{t \geq 0} \alpha^{t}\left[\begin{array}{c}
m+t \\
m
\end{array}\right]=\frac{1}{(\alpha)_{m+1}}
$$

Then

$$
B(a, b):=\sum_{t \geq 0}\left(\frac{p}{q}\right)^{t} A(a, b, t)=\frac{(p)_{a-1}}{(p)_{b}}
$$

and

$$
\begin{align*}
\mathbb{P}\left[I_{2}=i_{2}, J_{2}\right. & \left.=j_{2} \mid J_{1}=j_{1}\right]=\llbracket i_{2}=j_{1} \rrbracket \pi\left(j_{2}\right)+\frac{(p)_{j_{1}-1}}{(p)_{i_{2}}} \pi\left(i_{2}\right) \pi\left(j_{2}\right),  \tag{6.1}\\
\mathbb{P}\left[I_{2}\right. & \left.=i_{2} \mid J_{1}=j_{1}\right]=\llbracket i_{2}=j_{1} \rrbracket\left(1-q^{i_{2}-1}\right)+\frac{(p)_{j_{1}-1}}{(p)_{i_{2}}} p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) . \tag{6.2}
\end{align*}
$$

The transition matrix between $I_{1}$ and $I_{2}$ is given by

$$
\begin{align*}
& \mathbb{P}\left[I_{2}=i_{2} \mid I_{1}=i_{1}\right]=\sum_{j_{1}<i_{1}} \frac{\pi\left(j_{1}\right)}{1-P\left(i_{1}\right)} \mathbb{P}\left[I_{2}=i_{2} \mid J_{1}=j_{1}\right] \\
& \quad=\llbracket i_{2}<i_{1} \rrbracket \frac{\pi\left(i_{2}\right)}{1-P\left(i_{1}\right)}\left(1-q^{i_{2}-1}\right)+\sum_{j_{1}<i_{1}, j_{1} \leq i_{2}} \frac{\pi\left(j_{1}\right)}{1-P\left(i_{1}\right)} \frac{(p)_{j_{1}-1}}{(p)_{i_{2}}} \pi\left(i_{2}\right)\left(1-q^{i_{2}-1}\right) . \tag{6.3}
\end{align*}
$$

Let us first check that $\sum_{i_{2} \geq 2} \mathbb{P}\left[I_{2}=i_{2} \mid I_{1}=i_{1}\right]=1$. We have, using (A.8)

$$
\begin{aligned}
& \sum_{i_{2}=2}^{i_{1}-1} \frac{p q^{i_{2}-1}}{1-q^{i_{1}-1}}\left(1-q^{i_{2}-1}\right)+\sum_{i_{2} \geq 2} \sum_{j_{1}<i_{1}, j_{1} \leq i_{2}} \frac{p q^{j_{1}-1}}{1-q^{i_{1}-1}} \frac{(p)_{j_{1}-1}}{(p)_{i_{2}}} p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \\
& =\sum_{i_{2}=2}^{i_{1}-1} \frac{p q^{i_{2}-1}}{1-q^{i_{1}-1}}-\sum_{i_{2}=2}^{i_{1}-1} \frac{p q^{i_{2}-1}}{1-q^{i_{1}-1}} q^{i_{2}-1}+\sum_{i_{2} \geq 2} \frac{p}{1-q^{i_{1}-1}} \frac{1}{(p)_{i_{2}}} p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \\
& \quad+\sum_{j_{1}=2}^{i_{1}-1} \sum_{i_{2} \geq j_{1}} \frac{p q^{j_{1}-1}}{1-q^{i_{1}-1}} \frac{(p)_{j_{1}-1}}{(p)_{i_{2}}} p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \\
& =\frac{1}{1-q^{i_{1}-1}}\left[1-p-q^{i_{1}-1}\right]-\sum_{i_{2}=2}^{i_{1}-1} \frac{p q^{i_{2}-1}}{1-q^{i_{1}-1}} q^{i_{2}-1}+\frac{p}{1-q^{i_{1}-1}}+\frac{1}{1-q^{i_{1}-1}} \sum_{j_{1}=2}^{i_{1}-1} p q^{j_{1}-1} q^{j_{1}-1} \\
& =1 .
\end{aligned}
$$

Now we compute the stationary measure $\varphi(i)$ of this matrix. (see, for instance, (3)). We have

$$
\sum_{i_{1}>i_{2}} \varphi\left(i_{1}\right) \frac{p q^{i_{2}-1}}{1-q^{i_{1}-1}}\left(1-q^{i_{2}-1}\right)+\sum_{i_{1} \geq 2} \varphi\left(i_{1}\right) \sum_{j_{1}<i_{1}, j_{1} \leq i_{2}} \frac{p q^{j_{1}-1}}{1-q^{i_{1}-1}} \frac{(p)_{j_{1}-1}}{(p)_{i_{2}}} p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right)=\varphi\left(i_{2}\right)
$$

or, setting

$$
\psi(i)=\frac{\varphi(i)}{p q^{i-1}\left(1-q^{i-1}\right)},
$$

we have

$$
\sum_{i_{1}>i_{2}} p q^{i_{1}-1} \psi\left(i_{1}\right)+\sum_{i_{1} \geq 2} p q^{i_{1}-1} \psi\left(i_{1}\right) \sum_{j_{1}<i_{1}, j_{1} \leq i_{2}} p q^{j_{1}-1} \frac{(p)_{j_{1}-1}}{(p)_{i_{2}}}=\psi\left(i_{2}\right)
$$

After some algebra, we will find that $\psi(i)=$ constant is a solution of this equation. But it is probabilistically obvious: the stationary distribution is proportional to $p q^{i-1}\left(1-q^{i-1}\right)$. We have, setting $\psi \equiv 1$,

$$
\begin{aligned}
\sum_{i_{1}>i_{2}} p q^{i_{1}-1}+ & \sum_{i_{1} \geq 2} p q^{i_{1}-1} \sum_{j_{1}<i_{1}, j_{1} \leq i_{2}} p q^{j_{1}-1} \frac{(p)_{j_{1}-1}}{(p)_{i_{2}}}=q^{i_{2}}+\frac{1}{(p)_{i_{2}}} \sum_{j_{1}=1}^{i_{2}} p q^{j_{1}-1}(p)_{j_{1}-1} q^{j_{1}} \\
& =q^{i_{2}}+\frac{1}{(p)_{i_{2}}} p q \sum_{m \geq 0} q^{2 m}(p)_{m}-\frac{1}{(p)_{i_{2}}} p q q^{2 i_{2}}(p)_{i_{2}} \sum_{v \geq 0} q^{2 v}\left(p q^{i_{2}}\right)_{v} \\
& =q^{i_{2}}+\left(p q^{i_{2}}\right)_{\infty}+1-q^{i_{2}}-\left(p q^{i_{2}}\right)_{\infty}=1,
\end{aligned}
$$

as expected. (We used (A.5) twice.)
The stationary distribution is given by

$$
\begin{equation*}
f_{\text {stationary }}(i)=\frac{1+q}{q} p q^{i-1}\left(1-q^{i-1}\right) \tag{6.4}
\end{equation*}
$$

The stationary distribution generating function is given by

$$
G(z)=\frac{p^{2}(1+q) z^{2}}{(1-q z)\left(1-q^{2} z\right)}
$$

from which the stationary moments are easily derived:

$$
\begin{aligned}
\mathbb{E}(I) & =\frac{2+q}{1-q^{2}} \\
\mathbb{E}\left(I^{2}\right) & =\frac{q^{3}+4 q^{2}+5 q+4}{\left(1-q^{2}\right)^{2}}
\end{aligned}
$$

The stationay distribution of the end value is given by

$$
\frac{1+q}{q} \sum_{i_{1}=i_{2}+1}^{\infty} p q^{i_{1}-1}\left(1-q^{i_{1}-1}\right) p q^{i_{2}-1} /\left(1-q^{i_{1}-1}\right)=(1+q) p q^{2 i_{2}-2}, i_{2} \geq 1
$$

which sums correctly to 1 .
Another transition matrix of interest is given by

$$
\begin{equation*}
\mathbb{P}\left[J_{2}=j_{2} \mid J_{1}=j_{1}\right]=\pi\left(j_{2}\right) \llbracket j_{2}<j_{1} \rrbracket+\sum_{i_{2} \geq j_{1}, i_{2}>j_{2}} \frac{(p)_{j_{1}-1}}{(p)_{i_{2}}} \pi\left(i_{2}\right) \pi\left(j_{2}\right) \tag{6.5}
\end{equation*}
$$

## 7 Descent distributions

In this section, we use the Markov chain derived in the previous section to obtain the distribution of initial and end values of first and second descents. For the first descent initial value $I_{1}$, for example, with (A.9), we have

$$
f_{1}\left(i_{1}\right):=\mathbb{P}\left[I_{1}=i_{1}\right]=\sum_{j_{0}=1}^{i_{1}} p q^{j_{0}-1} \mathbb{P}\left[I_{1}=i_{1} \mid J_{0}=j_{0}\right]
$$

$$
\begin{align*}
& =p q^{i_{1}-1}\left(1-q^{i_{1}-1}\right)+\sum_{j_{0}=1}^{i_{1}} p q^{j_{0}-1} \frac{(p)_{j_{0}-1}}{(p)_{i_{1}}} p q^{i_{1}-1}\left(1-q^{i_{1}-1}\right) \\
& =p q^{i_{1}-1}\left(1-q^{i_{1}-1}\right) \frac{1}{(p)_{i_{1}}} \tag{7.1}
\end{align*}
$$

A graph of $f_{1}(i), q=0.7$ is given in Figure 1. It is easily checked that


Fig. 1: $f_{1}(i), q=0.7$

$$
\sum_{i=2}^{\infty} f_{1}(i)=1
$$

Indeed

$$
\sum_{i \geq 2} p q^{i-1}\left(1-q^{i-1}\right) \frac{1}{(p)_{i}}=1, \text { by (A.4) }
$$

A comparison between $f_{1}(i)$ and the stationary distribution $f_{\text {stationary }}(i), q=0.7$ is given in Figure 2. For the end height of the first descent, we have, with (A.9) and (A.10),

$$
\begin{align*}
\gamma_{1}\left(j_{1}\right) & :=\mathbb{P}\left[J_{1}=j_{1}\right]=\sum_{j_{0}>j_{1}} p q^{j_{0}-1} p q^{j_{1}-1}+\sum_{i_{1}>j_{1}} \sum_{j_{0}=1}^{i_{1}} p q^{j_{0}-1} \frac{(p)_{j_{0}-1}}{(p)_{i_{1}}} p q^{i_{1}-1} p q^{j_{1}-1} \\
& =p q^{j_{1}-1}\left[\frac{1}{(p)_{\infty}}-\frac{1}{(p)_{j_{1}}}\right] \tag{7.2}
\end{align*}
$$

Of course $\sum_{j \geq 1} \gamma_{1}(j)=1$, by (A.10).
Note that we also have, with (A.10),

$$
\gamma_{1}\left(j_{1}\right)=\sum_{j_{1}+1}^{\infty} f_{1}\left(i_{1}\right) \frac{\pi\left(j_{1}\right)}{1-P\left(i_{1}\right)}=\sum_{j_{1}+1}^{\infty} p q^{i_{1}-1}\left(1-q^{i_{1}-1}\right) \frac{1}{(p)_{i_{1}}} \frac{p q^{j_{1}-1}}{1-q^{i_{1}-1}}=p q^{j_{1}-1}\left[\frac{1}{(p)_{\infty}}-\frac{1}{(p)_{j_{1}}}\right]
$$



Fig. 2: Comparison between $f_{1}(i)$ and $f_{\text {stationary }}(i), q=0.7$

The distribution of the second descent $\left(I_{2}, J_{2}\right)$ is given by the square of the Markov matrix, i.e.

$$
\gamma_{2}\left(i_{2}, j_{2}\right)=\sum_{j_{0} \geq 1} p q^{j_{0}-1} \sum_{j_{1}} \mathbb{P}\left[J_{1}=j_{1} \mid J_{0}=j_{0}\right] \mathbb{P}\left[I_{2}=i_{2}, J_{2}=j_{2} \mid J_{1}=j_{1}\right]
$$

The distribution of the next descents is related to successive powers of the transition matrix.
A compact form for $\gamma_{2}\left(i_{2}, j_{2}\right)$ is given below.
Now, the distribution of the initial value of the second descent $I_{2}$ is given by (we use (A.3), (A.9))

$$
\begin{aligned}
f_{2}\left(i_{2}\right) & :=\mathbb{P}\left[I_{2}=i_{2}\right]=\sum_{i_{1}=2}^{\infty} f_{1}\left(i_{1}\right) \mathbb{P}\left[I_{2}=i_{2} \mid I_{1}=i_{1}\right]=\sum_{i_{1}=2}^{\infty} p q^{i_{1}-1}\left(1-q^{i_{1}-1}\right) \frac{1}{(p)_{i_{1}}} \mathbb{P}\left[I_{2}=i_{2} \mid I_{1}=i_{1}\right] \\
& =\sum_{i_{1}>i_{2}} p q^{i_{1}-1}\left(1-q^{i_{1}-1}\right) \frac{1}{(p)_{i_{1}}} \frac{p q^{i_{2}-1}}{1-q^{i_{1}-1}}\left(1-q^{i_{2}-1}\right) \\
& +\sum_{i_{1} \geq 2} p q^{i_{1}-1}\left(1-q^{i_{1}-1}\right) \frac{1}{(p)_{i_{1}}} \sum_{j_{1}<i_{1}, j_{1} \leq i_{2}} \frac{p q^{j_{1}-1}}{1-q^{i_{1}-1}} \frac{(p)_{j_{1}-1}}{(p)_{i_{2}}} p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \\
& =p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}-1}\left(1-p q^{i_{2}-1}\right)}\left[\frac{1}{\left(p q^{i_{2}}\right)_{\infty}}-1\right] \\
& +p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}}} \sum_{j_{1}=1}^{i_{2}} p q^{j_{1}-1}(p)_{j_{1}-1} \sum_{i_{1}>j_{1}} p q^{i_{1}-1} \frac{1}{(p)_{i_{1}}}
\end{aligned}
$$

$$
\begin{aligned}
& =p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}-1}\left(1-p q^{i_{2}-1}\right)}\left[\frac{1}{\left(p q^{i_{2}}\right)_{\infty}}-1\right] \\
& +p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}}} \sum_{j_{1}=1}^{i_{2}} p q^{j_{1}-1} \frac{1}{1-p q^{j_{1}-1}}\left[\frac{1}{\left(p q^{j_{1}}\right)_{\infty}}-1\right] \\
& =p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}}}\left[\frac{1}{\left(p q^{\left.i_{2}\right)_{\infty}}-1\right]}\right. \\
& +p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}}} \sum_{j_{1}=1}^{i_{2}} p q^{j_{1}-1}\left[\frac{1}{\left(p q^{j_{1}-1}\right)_{\infty}}-\frac{1}{1-p q^{j_{1}-1}}\right] \\
& =p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right)\left[\frac{1}{(p)_{\infty}}-\frac{1}{(p)_{i_{2}}}\right] \\
& +p^{2} q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}}} \frac{1}{(p)_{\infty}} \sum_{j_{1}=0}^{i_{2}-1} q^{j_{1}}(p)_{j_{1}}-p^{2} q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}}} \sum_{j_{1}=0}^{i_{2}-1} \frac{q^{j_{1}}}{1-p q^{j_{1}}} \\
& =p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right)\left[\frac{1}{(p)_{\infty}}-\frac{1}{(p)_{i_{2}}}\right] \\
& +p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}}} \frac{1}{(p)_{\infty}}\left[1-(p)_{i_{2}}\right]-p^{2} q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}}} \sum_{j_{1}=0}^{i_{2}-1} \frac{q^{j_{1}}}{1-p q^{j_{1}}} \\
& =p q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}}}\left[\frac{1}{(p)_{\infty}}-1\right]-p^{2} q^{i_{2}-1}\left(1-q^{i_{2}-1}\right) \frac{1}{(p)_{i_{2}}} \sum_{j_{1}=0}^{i_{2}-1} \frac{q^{j_{1}}}{1-p q^{j_{1}}} .
\end{aligned}
$$

Note that this gives an explicit expression for $\gamma_{2}\left(i_{2}, j_{2}\right)$ :

$$
\gamma_{2}\left(i_{2}, j_{2}\right)=f_{2}\left(i_{2}\right) \frac{\pi\left(j_{2}\right)}{1-P\left(i_{2}\right)}
$$

A comparison between $f_{2}(i)$ and $f_{\text {stationary }}(i), q=0.7$ is given in Figure 3. The convergence to the stationary distribution is quite fast.

Let us check that $\sum_{i=2}^{\infty} f_{2}(i)=1$. We have, using (A.8) and (A.1)

$$
\begin{aligned}
& \sum_{i=2}^{\infty}\left[p q^{i-1}\left(1-q^{i-1}\right) \frac{1}{(p)_{i}}\left[\frac{1}{(p)_{\infty}}-1\right]-p^{2} q^{i-1}\left(1-q^{i-1}\right) \frac{1}{(p)_{i}} \sum_{j_{1}=0}^{i-1} \frac{q^{j_{1}}}{1-p q^{j_{1}}}\right] \\
& =\frac{1}{(p)_{\infty}}-1-\sum_{j_{1} \geq 1} p^{2} \frac{q^{j_{1}}}{1-p q^{j_{1}}} \sum_{i \geq j_{1}+1} q^{i-1}\left(1-q^{i-1}\right) \frac{1}{(p)_{i}}-p^{2} \frac{1}{1-p} \sum_{i \geq 2} q^{i-1}\left(1-q^{i-1}\right) \frac{1}{(p)_{i}} \\
& =\frac{1}{(p)_{\infty}}-1-\sum_{j_{1} \geq 1} p^{2} \frac{q^{j_{1}}}{1-p q^{j_{1}}} \frac{q^{j_{1}}}{p(p)_{j_{1}}}-p^{2} \frac{1}{1-p} \frac{1}{p}=1 .
\end{aligned}
$$

## 8 The moments of descent parameters

The first moments of the first and second descents initial values are analyzed here by intensive use of some combinatorial identities.


$$
\circ: f_{2}(i)
$$

line : $f_{\text {stationary }}(i)$
Fig. 3: Comparison between $f_{2}(i)$ and $f_{\text {stationary }}(i), q=0.7$
We derive the mean of the initial value of the first descent $I_{1}$,

$$
\begin{align*}
\mathbb{E}\left(I_{1}\right) & =\sum_{i=2}^{\infty} f_{1}(i) i=\sum_{i \geq 2} p q^{i-1}\left(1-q^{i-1}\right) \frac{1}{(p)_{i}} i \\
& =p \sum_{i \geq 1} \frac{q^{i-1} i}{(p q)_{i}}-p \sum_{i \geq 1} \frac{q^{2 i-1} i}{(p q)_{i}}+p \sum_{i \geq 1} \frac{q^{i}}{(p)_{i+1}}\left(1-q^{i}\right)  \tag{8.1}\\
& =p \sum_{i \geq 1} \frac{q^{i-1} i}{(p q)_{i}}-p \sum_{i \geq 1} \frac{q^{2 i-1} i}{(p q)_{i}}+1 \text { by (A.4). }
\end{align*}
$$

This is Theorem 3 in (5). However, we want to show now independently that this coincides with closed form obtained earlier as (3.4).

$$
\begin{aligned}
\mathbb{E}\left(I_{1}\right) & =p \sum_{i \geq 0} \frac{q^{i}}{(p)_{i+1}}\left(1-q^{i}\right)(i+1) \\
& =p \sum_{i \geq 0} \sum_{h=0}^{i} \frac{q^{i}}{(p)_{i+1}}\left(1-q^{i}\right) \\
& =\frac{p}{q} \sum_{h \geq 0} \sum_{i \geq h} \frac{q^{i}}{(p q)_{i}}-\frac{p}{q} \sum_{h \geq 0} \sum_{i \geq h} \frac{q^{2 i}}{(p q)_{i}}
\end{aligned}
$$

$$
=\frac{p}{q} \sum_{h \geq 0} \frac{q^{h}}{(p q)_{h}} \sum_{i \geq 0} \frac{q^{i}}{\left(p q^{h+1}\right)_{i}}-\frac{p}{q} \sum_{h \geq 0} \frac{q^{2 h}}{(p q)_{h}} \sum_{i \geq 0} \frac{q^{2 i}}{\left(p q^{h+1}\right)_{i}}
$$

Now use (A.3):

$$
\sum_{i \geq 0} \frac{q^{i}}{\left(p q^{h+1}\right)_{i}}=\frac{1}{p q^{h}\left(p q^{h+1}\right)_{\infty}}+1-\frac{1}{p q^{h}}
$$

We need also this (use(A.1)):

$$
\sum_{i \geq 0} \frac{q^{2 i}}{\left(p q^{h+1}\right)_{i}}=\frac{1}{p q^{2 h}} \frac{1}{\left(p q^{h+1}\right)_{\infty}}-\frac{1}{p q^{2 h}}-\frac{1}{p q^{h-1}}+1
$$

We now plug these two result in:

$$
\begin{aligned}
\mathbb{E}\left(I_{1}\right) & =\frac{p}{q} \sum_{h \geq 0} \frac{q^{h}}{(p q)_{h}} \sum_{i \geq 0} \frac{q^{i}}{\left(p q^{h+1}\right)_{i}}-\frac{p}{q} \sum_{h \geq 0} \frac{q^{2 h}}{(p q)_{h}} \sum_{i \geq 0} \frac{q^{2 i}}{\left(p q^{h+1}\right)_{i}} \\
& =\frac{p}{q} \sum_{h \geq 0} \frac{q^{h}}{(p q)_{h}}\left[\frac{1}{p q^{h}\left(p q^{h+1}\right)_{\infty}}+1-\frac{1}{p q^{h}}\right]-\frac{p}{q} \sum_{h \geq 0} \frac{q^{2 h}}{(p q)_{h}}\left[\frac{1}{p q^{2 h}} \frac{1}{\left(p q^{h+1}\right)_{\infty}}-\frac{1}{p q^{2 h}}-\frac{1}{p q^{h-1}}+1\right] \\
& =\frac{p}{q} \sum_{h \geq 0} \frac{1}{(p q)_{h}}\left[\frac{1}{p\left(p q^{h+1}\right)_{\infty}}+q^{h}-\frac{1}{p}\right]-\frac{p}{q} \sum_{h \geq 0} \frac{1}{(p q)_{h}}\left[\frac{1}{p\left(p q^{h+1}\right)_{\infty}}-\frac{1}{p}-\frac{q^{h+1}}{p}+q^{2 h}\right] \\
& =\frac{p}{q} \sum_{h \geq 0} \frac{1}{(p q)_{h}}\left[q^{h}+\frac{q^{h+1}}{p}-q^{2 h}\right] \\
& =\frac{p}{q} \sum_{h \geq 0} \frac{1}{(p q)_{h}}\left[q^{h}-q^{2 h}\right]+\sum_{h \geq 0} \frac{q^{h}}{(p q)_{h}} \\
& =\frac{p}{q} \sum_{h \geq 0} \frac{q^{h}}{(p q)_{h}}-\frac{p}{q} \sum_{h \geq 0} \frac{q^{2 h}}{(p q)_{h}}+\sum_{h \geq 0} \frac{q^{h}}{(p q)_{h}} \\
& =\frac{1}{q} \sum_{h \geq 0} \frac{q^{h}}{(p q)_{h}}-\frac{p}{q} \sum_{h \geq 0} \frac{q^{2 h}}{(p q)_{h}} \\
& =\frac{1}{q}\left[\frac{1}{p(p q)_{\infty}}+1-\frac{1}{p}\right]-\frac{p}{q}\left[\frac{1}{p} \frac{1}{(p q)_{\infty}}-\frac{1}{p}-\frac{q}{p}+1\right] \\
& =\frac{1}{q}\left[\frac{q}{p(p q)_{\infty}}-\frac{q}{p}+2 q\right] \\
& =\frac{1}{p(p q)_{\infty}}-\frac{1}{p}+2 .
\end{aligned}
$$

Since

$$
\frac{1}{p(p q)_{\infty}}-\frac{1}{p}+2=1-\frac{q}{p}+\frac{q}{p(p)_{\infty}}
$$

we established the formula (3.4)

$$
\mathbb{E}\left(I_{1}\right)=1+\frac{q}{p}\left(-1+\frac{1}{(p)_{\infty}}\right)
$$

More generally higher moments are given by

$$
\mathbb{E}\left(I_{1}^{k}\right)=\sum_{h \geq 1} p q^{h}\left(1-q^{h}\right)(h+1)^{k}+\sum_{i_{1}=2}^{\infty} \sum_{j_{0}=1}^{i_{1}} p q^{j_{0}-1}(p)_{j_{0}-1} \frac{p q^{i_{1}-1}}{(p)_{i_{1}}}\left(1-q^{i_{1}-1}\right) i_{1}^{k}
$$

Let us look at the modified second moment of $I_{1}$ (again we use (A.3), (A.1) for simplifications):

$$
\begin{aligned}
\mathbb{E} \mathbb{E} & =\sum_{i=2}^{\infty} f_{1}(i) i(i+1) / 2=p \sum_{i \geq 0} \frac{q^{i}}{(p)_{i+1}}\left(1-q^{i}\right)(i+1)(i+2) / 2 \\
& =p \sum_{0 \leq k \leq h} \sum_{i \geq h} \frac{q^{i}}{(p)_{i+1}}\left(1-q^{i}\right) \\
& =\frac{1}{q} \sum_{0 \leq k \leq h} \frac{q^{h}}{(p q)_{h}}-\frac{p}{q} \sum_{0 \leq k \leq h} \frac{q^{2 h}}{(p q)_{h}} \\
& =\frac{1}{q} \sum_{k \geq 0} \frac{q^{k}}{(p q)_{k}} \sum_{h \geq 0} \frac{q^{h}}{\left(p q^{k+1}\right)_{h}}-\frac{p}{q} \sum_{k \geq 0} \frac{q^{2 k}}{(p q)_{k}} \sum_{h \geq 0} \frac{q^{2 h}}{\left(p q^{k+1}\right)_{h}} \\
& =\frac{1}{q} \sum_{k \geq 0} \frac{q^{k}}{(p q)_{k}}\left[\frac{1}{p q^{k}\left(p q^{k+1}\right)_{\infty}}+1-\frac{1}{p q^{k}}\right]-\frac{p}{q} \sum_{k \geq 0} \frac{q^{2 k}}{(p q)_{k}}\left[\frac{1}{p q^{2 k}} \frac{1}{\left(p q^{k+1}\right)_{\infty}}-\frac{1}{p q^{2 k}}-\frac{1}{p q^{k-1}}+1\right] \\
& =\frac{1}{q} \sum_{k \geq 0} \frac{1}{(p q)_{k}}\left[\frac{1}{p\left(p q^{k+1}\right)_{\infty}}+q^{k}-\frac{1}{p}\right]-\frac{1}{q} \sum_{k \geq 0} \frac{1}{(p q)_{k}}\left[\frac{1}{\left(p q^{k+1}\right)_{\infty}}-1-q^{k+1}+p q^{2 k}\right] \\
& =\frac{1}{q} \sum_{k \geq 0} \frac{1}{(p q)_{k}}\left[\frac{1}{p\left(p q^{k+1}\right)_{\infty}}+q^{k}-\frac{1}{p}-\frac{1}{\left(p q^{k+1}\right)_{\infty}}+1+q^{k+1}-p q^{2 k}\right] \\
& =\frac{1}{q} \sum_{k \geq 0} \frac{1}{(p q)_{k}}\left[\frac{q}{p\left(p q^{k+1}\right)_{\infty}}-\frac{q}{p}+q^{k}(1+q)-p q^{2 k}\right] \\
& \left.=\frac{1}{p} \sum_{k \geq 0} \frac{1}{(p q)_{k}}\left[\frac{1}{\left(p q^{k+1}\right)_{\infty}}-1\right]+\frac{1+q}{q} \sum_{k \geq 0} \frac{q^{k}}{(p q)_{k}}-\frac{p}{q} \sum_{k \geq 0}^{(p q)_{k}} \frac{q^{2 k}}{(p l}\right]-\frac{p}{q}\left[\frac{1}{p} \frac{1}{(p q)_{\infty}}-\frac{1}{p}-\frac{q}{p}+1\right] \\
& =\frac{1}{p} \sum_{k \geq 0}\left[\frac{1}{(p q)_{\infty}}-\frac{1}{(p q)_{k}}\right]+\frac{1+q}{q}\left[\frac{1}{p(p q)_{\infty}}+1-\frac{1}{p}\right] \\
& =\frac{1}{p} \sum_{k \geq 0}\left[\frac{1}{(p q)_{\infty}}-\frac{1}{(p q)_{k}}\right]+\frac{2}{p(p q)_{\infty}}+3-\frac{2}{p} \\
& =\frac{1}{p} \lim _{t \rightarrow 1}\left[\frac{1}{(p q)_{\infty}} \frac{1}{1-t}-\sum_{k \geq 0} \frac{t^{k}}{(p q)_{k}}\right]+\frac{2}{p(p q)_{\infty}}+3-\frac{2}{p} .
\end{aligned}
$$

To compute this limit, we use (A.7):

$$
\lim _{t \rightarrow 1}\left[\frac{1}{(p q)_{\infty}} \frac{1}{1-t}-\sum_{k \geq 0} \frac{t^{k}}{(p q)_{k}}\right]
$$

$$
\begin{aligned}
& =\lim _{t \rightarrow 1}\left[\frac{1}{(p q)_{\infty}} \frac{1}{1-t}-\frac{(q)_{\infty}}{(p q)_{\infty}(t)_{\infty}} \sum_{m \geq 0} \frac{(p)_{m}(t)_{m} q^{m}}{(q)_{m}}\right] \\
& =\lim _{t \rightarrow 1}\left[\frac{1}{(p q)_{\infty}} \frac{1}{1-t}-\frac{(q)_{\infty}}{(p q)_{\infty}(t)_{\infty}}-\frac{(q)_{\infty}}{(p q)_{\infty}(t)_{\infty}} \sum_{m \geq 1} \frac{(p)_{m}(t)_{m} q^{m}}{(q)_{m}}\right] \\
& =\lim _{t \rightarrow 1}\left[\frac{1}{(p q)_{\infty}} \frac{1}{1-t}-\frac{(q)_{\infty}}{(p q)_{\infty}(1-t)(q t)_{\infty}}\right]-\frac{(q)_{\infty}}{(p q)_{\infty}(q)_{\infty}} \sum_{m \geq 1} \frac{(p)_{m}(q)_{m-1} q^{m}}{(q)_{m}} \\
& =\frac{1}{(p q)_{\infty}} \lim _{t \rightarrow 1}\left[\frac{1}{1-t}-\frac{(q)_{\infty}}{(1-t)(q t)_{\infty}}\right]-\frac{1}{(p q)_{\infty}} \sum_{m \geq 1} \frac{(p)_{m}(q)_{m-1} q^{m}}{(q)_{m}} \\
& =\frac{1}{(p q)_{\infty}} \lim _{t \rightarrow 1} \frac{(q t)_{\infty}-(q)_{\infty}}{(1-t)(q t)_{\infty}}-\frac{1}{(p q)_{\infty}} \sum_{m \geq 1} \frac{(p)_{m}(q)_{m-1} q^{m}}{(q)_{m}} \\
& =\frac{1}{(p q)_{\infty}} \lim _{t \rightarrow 1} \frac{\left.\frac{d}{d t}(q t)_{\infty}\right|_{t=1}(t-1)}{(1-t)(q t)_{\infty}}-\frac{1}{(p q)_{\infty}} \sum_{m \geq 1} \frac{(p)_{m}(q)_{m-1} q^{m}}{(q)_{m}} \\
& =-\left.\frac{1}{(p q)_{\infty}(q)_{\infty}} \frac{d}{d t}(q t)_{\infty}\right|_{t=1}-\frac{1}{(p q)_{\infty}} \sum_{m \geq 1} \frac{(p)_{m}(q)_{m-1} q^{m}}{(q)_{m}} \\
& =\frac{1}{(p q)_{\infty}} \sum_{k \geq 1} \frac{q^{k}}{1-q^{k}}-\frac{1}{(p q)_{\infty}} \sum_{m \geq 1} \frac{(p)_{m} q^{m}}{1-q^{m}} \\
& =\frac{1}{(p q)_{\infty}} \sum_{k \geq 1} \frac{\left[1-(p)_{k}\right] q^{k}}{1-q^{k}} .
\end{aligned}
$$

So finally:

$$
\mathbb{E} \mathbb{E}=\frac{1}{p(p q)_{\infty}} \sum_{k \geq 1} \frac{\left[1-(p)_{k}\right] q^{k}}{1-q^{k}}+\frac{2}{p(p q)_{\infty}}+3-\frac{2}{p}
$$

From this, the variance can we stated as:

$$
2 \mathbb{E} \mathbb{E}-2 \mathbb{E}\left(I_{1}\right)-\mathbb{E}^{2}\left(I_{1}\right)
$$

Now we turn to the second descent. The mean $\mathbb{E}\left(I_{2}\right)$ of the initial value of the second descent is given by (we use (3.4))

$$
\begin{aligned}
\mathbb{E}\left(I_{2}\right) & =\sum_{i=2}^{\infty} f_{2}(i) i=p\left[\frac{1}{(p)_{\infty}}-1\right] \sum_{i \geq 1} i q^{i-1}\left(1-q^{i-1}\right) \frac{1}{(p)_{i}}-\sum_{i \geq 1} i p^{2} q^{i-1}\left(1-q^{i-1}\right) \frac{1}{(p)_{i}} \sum_{j=0}^{i-1} \frac{q^{j}}{1-p q^{j}} \\
& =p\left[\frac{1}{(p)_{\infty}}-1\right] \sum_{i \geq 0}(i+1) q^{i}\left(1-q^{i}\right) \frac{1}{(p)_{i+1}}-p^{2} \sum_{j \geq 0} \frac{q^{j}}{1-p q^{j}} \sum_{i \geq j}(i+1) q^{i}\left(1-q^{i}\right) \frac{1}{(p)_{i+1}} \\
& =\left[\frac{1}{(p)_{\infty}}-1\right]\left[1-\frac{q}{p}+\frac{q}{p(p)_{\infty}}\right]-p \sum_{j \geq 0} \frac{p q^{j}}{1-p q^{j}} \sum_{i \geq j}(i+1) q^{i}\left(1-q^{i}\right) \frac{1}{(p)_{i+1}} \\
& =\left[\frac{1}{(p)_{\infty}}-1-\sum_{j \geq 0} \frac{p q^{j}}{1-p q^{j}}\right]\left[1-\frac{q}{p}+\frac{q}{p(p)_{\infty}}\right]+p \sum_{j \geq 0} \frac{p q^{j}}{1-p q^{j}} \sum_{0 \leq i<j}(i+1) q^{i}\left(1-q^{i}\right) \frac{1}{(p)_{i+1}} .
\end{aligned}
$$

The last inner sum can be computed:

$$
\begin{aligned}
S_{0} & :=\sum_{0 \leq i<j}(i+1) q^{i}\left(1-q^{i}\right) \frac{1}{(p)_{i+1}}=\sum_{0 \leq i<j} \sum_{0 \leq h \leq i} q^{i}\left(1-q^{i}\right) \frac{1}{(p)_{i+1}} \\
& =\sum_{0 \leq h \leq i<j} \frac{q^{i}}{(p)_{i+1}}-\sum_{0 \leq h \leq i<j} \frac{q^{2 i}}{(p)_{i+1}} \\
& =\sum_{0 \leq h<j} \sum_{0 \leq i<j-h} \frac{q^{i+h}}{(p)_{i+h+1}}-\sum_{0 \leq h<j} \sum_{0 \leq i<j-h} \frac{q^{2 i+2 h}}{(p)_{i+h+1}} \\
& =\sum_{0 \leq h<j} \frac{q^{h}}{(p)_{h+1}} \sum_{0 \leq i<j-h} \frac{q^{i}}{\left(p q^{h+1}\right)_{i}}-\sum_{0 \leq h<j} \frac{q^{2 h}}{(p)_{h+1}} \sum_{0 \leq i<j-h} \frac{q^{2 i}}{\left(p q^{h+1}\right)_{i}} .
\end{aligned}
$$

We do two auxiliary calculation (with (A.3)):

$$
\begin{aligned}
S_{1} & :=\sum_{0 \leq i<I} \frac{q^{i}}{\left(p q^{h+1}\right)_{i}} \\
& =\sum_{i \geq 0} \frac{q^{i}}{\left(p q^{h+1}\right)_{i}}-\sum_{i \geq I} \frac{q^{i}}{\left(p q^{h+1}\right)_{i}} \\
& =\frac{1}{p q^{h}\left(p q^{h+1}\right)_{\infty}}+1-\frac{1}{p q^{h}}-\frac{q^{I}}{\left(p q^{h+1}\right)_{I}} \sum_{i \geq 0} \frac{q^{i}}{\left(p q^{h+I+1}\right)_{i}} \\
& =\frac{1}{p q^{h}\left(p q^{h+1}\right)_{\infty}}+1-\frac{1}{p q^{h}}-\frac{q^{I}}{\left(p q^{h+1}\right)_{I}}\left[\frac{1}{p q^{h+I}\left(p q^{h+I+1}\right)_{\infty}}+1-\frac{1}{p q^{h+I}}\right] \\
& =1-\frac{1}{p q^{h}}-\frac{q^{I}}{\left(p q^{h+1}\right)_{I}}+\frac{1}{p q^{h}\left(p q^{h+1}\right)_{I}},
\end{aligned}
$$

and with (A.1):

$$
\begin{aligned}
S_{2} & :=\sum_{0 \leq i<I} \frac{q^{2 i}}{\left(p q^{h+1}\right)_{i}} \\
& =\sum_{i \geq 0} \frac{q^{2 i}}{\left(p q^{h+1}\right)_{i}}-\sum_{i \geq I} \frac{q^{2 i}}{\left(p q^{h+1}\right)_{i}} \\
& =\frac{1}{p q^{2 h}\left(p q^{h+1}\right)_{\infty}}-\frac{1}{p q^{2 h}}-\frac{1}{p q^{h-1}}+1-\frac{q^{2 I}}{\left(p q^{h+1}\right)_{I}} \sum_{i \geq 0} \frac{q^{2 i}}{\left(p q^{h+I+1}\right)_{i}} \\
& =\frac{1}{p q^{2 h}\left(p q^{h+1}\right)_{\infty}}-\frac{1}{p q^{2 h}}-\frac{1}{p q^{h-1}}+1-\frac{q^{2 I}}{\left(p q^{h+1}\right)_{I}}\left[\frac{1}{p q^{2 h+2 I}\left(p q^{h+I+1}\right)_{\infty}}-\frac{1}{p q^{2 h+2 I}}-\frac{1}{p q^{h+I-1}}+1\right] \\
& =-\frac{1}{p q^{2 h}}-\frac{1}{p q^{h-1}}+1+\frac{1}{p q^{2 h}\left(p q^{h+1}\right)_{I}}+\frac{q^{I}}{p q^{h-1}\left(p q^{h+1}\right)_{I}}-\frac{q^{2 I}}{\left(p q^{h+1}\right)_{I}} .
\end{aligned}
$$

Therefore

$$
\begin{aligned}
S_{0} & =\sum_{0 \leq h<j} \frac{q^{h}}{(p)_{h+1}} \sum_{0 \leq i<j-h} \frac{q^{i}}{\left(p q^{h+1}\right)_{i}}-\sum_{0 \leq h<j} \frac{q^{2 h}}{(p)_{h+1}} \sum_{0 \leq i<j-h} \frac{q^{2 i}}{\left(p q^{h+1}\right)_{i}} \\
& =\sum_{0 \leq h<j} \frac{q^{h}}{(p)_{h+1}}\left[1-\frac{1}{p q^{h}}-\frac{q^{j-h}}{\left(p q^{h+1}\right)_{j-h}}+\frac{1}{p q^{h}\left(p q^{h+1}\right)_{j-h}}\right] \\
& -\sum_{0 \leq h<j} \frac{q^{2 h}}{(p)_{h+1}}\left[-\frac{1}{p q^{2 h}}-\frac{1}{p q^{h-1}}+1+\frac{1}{p q^{2 h}\left(p q^{h+1}\right)_{j-h}}+\frac{q^{j-h}}{p q^{h-1}\left(p q^{h+1}\right)_{j-h}}-\frac{q^{2(j-h)}}{\left(p q^{h+1}\right)_{j-h}}\right] \\
& =\sum_{0 \leq h<j}\left[\frac{q^{h}}{(p)_{h+1}}-\frac{1}{p(p)_{h+1}}-\frac{q^{j}}{(p)_{j+1}}+\frac{1}{p(p)_{j+1}}\right] \\
& -\sum_{0 \leq h<j}\left[-\frac{1}{p(p)_{h+1}}-\frac{q^{h+1}}{p(p)_{h+1}}+\frac{q^{2 h}}{(p)_{h+1}}+\frac{1}{p(p)_{j+1}}+\frac{q^{j+1}}{p(p)_{j+1}}-\frac{q^{2 j}}{(p)_{j+1}}\right] \\
& =\sum_{0 \leq h<j}\left[\frac{q^{h}}{p(p)_{h+1}}-\frac{q^{j}}{p(p)_{j+1}}-\frac{q^{2 h}}{(p)_{h+1}}+\frac{q^{2 j}}{(p)_{j+1}}\right] .
\end{aligned}
$$

Recalling the expressions for $S_{1}$ and $S_{2}$ that we just derived, we get

$$
\begin{aligned}
S_{0} & =\sum_{0 \leq h<j}\left[\frac{q^{h}}{p q(p q)_{h}}-\frac{q^{2 h}}{q(p q)_{h}}\right]-\frac{j q^{j}}{p(p)_{j+1}}+\frac{j q^{2 j}}{(p)_{j+1}} \\
& =\frac{1}{p q}\left[1-\frac{1}{p}-\frac{q^{j}}{(p q)_{j}}+\frac{1}{p(p q)_{j}}\right]-\frac{1}{q}\left[-\frac{2 q}{p}+\frac{1}{p(p q)_{j}}+\frac{q^{j+1}}{p(p q)_{j}}-\frac{q^{2 j}}{(p)_{j}}\right]-\frac{j q^{j}}{p(p)_{j+1}}+\frac{j q^{2 j}}{(p)_{j+1}} \\
& =\frac{1}{p q}-\frac{1}{p^{2} q}-\frac{q^{j}}{p q(p q)_{j}}+\frac{1}{p^{2} q(p q)_{j}}+\frac{2}{p}-\frac{1}{p q(p q)_{j}}-\frac{q^{j}}{p(p q)_{j}}+\frac{q^{2 j}}{q(p)_{j}}-\frac{j q^{j}}{p(p)_{j+1}}+\frac{j q^{2 j}}{(p)_{j+1}} \\
& =\frac{2}{p}-\frac{1}{p^{2}}+\frac{q}{p^{2}(p)_{j+1}}-\frac{(1+q) q^{j}}{p q(p)_{j+1}}+\frac{q^{2 j}}{q(p)_{j}}-\frac{j q^{j}}{p(p)_{j+1}}+\frac{j q^{2 j}}{(p)_{j+1}} .
\end{aligned}
$$

Therefore we found a first expression for the mean:

$$
\begin{aligned}
\mathbb{E}\left(I_{2}\right) & =\frac{1}{p}\left[\frac{1}{(p)_{\infty}}-1-\sum_{j \geq 0} \frac{p q^{j}}{1-p q^{j}}\right]\left[2 p-1+\frac{q}{(p)_{\infty}}\right] \\
& +\frac{1}{p} \sum_{j \geq 0} \frac{p q^{j}}{1-p q^{j}}\left[2 p-1+\frac{q}{(p)_{j+1}}-\frac{p(1+q) q^{j}}{q(p)_{j+1}}+\frac{p^{2} q^{2 j}}{q(p)_{j}}-\frac{p j q^{j}}{(p)_{j+1}}+\frac{j p^{2} q^{2 j}}{(p)_{j+1}}\right] \\
& =\frac{1}{p}\left[\frac{1}{(p)_{\infty}}-1\right]\left[2 p-1+\frac{q}{(p)_{\infty}}\right] \\
& +\frac{1}{p} \sum_{j \geq 0} \frac{p q^{j}}{1-p q^{j}}\left[\frac{q}{(p)_{j+1}}-\frac{q}{(p)_{\infty}}-\frac{p(1+q) q^{j}}{q(p)_{j+1}}+\frac{p^{2} q^{2 j}}{q(p)_{j}}-\frac{p j q^{j}}{(p)_{j+1}}+\frac{j p^{2} q^{2 j}}{(p)_{j+1}}\right]
\end{aligned}
$$

$$
\begin{aligned}
& =\frac{1}{p}\left[\frac{1}{(p)_{\infty}}-1\right]\left[2 p-1+\frac{q}{(p)_{\infty}}\right] \\
& +\frac{1}{p} \sum_{j \geq 0} \frac{p q^{j}}{1-p q^{j}}\left[\frac{q}{(p)_{j+1}}-\frac{q}{(p)_{\infty}}-\frac{p(1+q) q^{j}}{q(p)_{j+1}}-\frac{p j q^{j}}{(p)_{j+1}}+\frac{j p^{2} q^{2 j}}{(p)_{j+1}}\right]+\frac{1}{p} \sum_{j \geq 0} \frac{p^{3} q^{3 j}}{q(p)_{j+1}}
\end{aligned}
$$

The last sum can be simplified; we will use Heine with $t=p q^{h}, a=0, c=q, b=q$ :

$$
\begin{aligned}
\sum_{i \geq 0} \frac{q^{3 i}}{\left(p q^{h+1}\right)_{i}} & =\frac{(q)_{2}}{\left(p q^{h+1}\right)_{\infty}} \sum_{m \geq 0} \frac{\left(p q^{h}\right)_{m}\left(q^{3}\right)_{m}}{(q)_{m}} q^{m} \\
& =\frac{(q)_{2}}{\left(p q^{h+1}\right)_{\infty}} \frac{\left(p q^{h}\right)_{\infty}\left(q^{4}\right)_{\infty}}{(q)_{\infty}} \sum_{m \geq 0} \frac{(q)_{m}}{(q)_{m}\left(q^{4}\right)_{m}}\left(p q^{h}\right)^{m} \\
& =\frac{1-p q^{h}}{1-q^{3}} \sum_{m \geq 0} \frac{1}{\left(q^{4}\right)_{m}}\left(p q^{h}\right)^{m} \\
& =\frac{\left(1-p q^{h}\right)(q)_{3}}{\left(p q^{h}\right)^{3}} \sum_{m \geq 3} \frac{1}{(q)_{m}}\left(p q^{h}\right)^{m} \\
& =\frac{\left(1-p q^{h}\right)(q)_{3}}{\left(p q^{h}\right)^{3}}\left[\frac{1}{\left(p q^{h}\right)_{\infty}}-1-q^{h}-\frac{\left(p q^{h}\right)^{2}}{(q)_{2}}\right] \\
& =\frac{(q)_{3}}{\left(p q^{h}\right)^{3}} \frac{1}{\left(p q^{h+1}\right)_{\infty}}-\frac{\left(1-p q^{h}\right)(q)_{3}}{p^{3} q^{3 h}}-\frac{\left(1-p q^{h}\right)(q)_{3}}{p^{3} q^{2 h}}-\frac{\left(1-p q^{h}\right)\left(1-q^{3}\right)}{p q^{h}}
\end{aligned}
$$

From this we derive (set $h=-1$ )

$$
\begin{aligned}
\sum_{j \geq 0} \frac{p^{3} q^{3 j}}{q(p)_{j+1}} & =\frac{p^{3}}{q}\left[\frac{(q)_{3}}{(p / q)^{3}} \frac{1}{(p)_{\infty}}-\frac{(1-p / q)(q)_{3}}{p^{3} q^{-3}}-\frac{(1-p / q)(q)_{3}}{p^{3} q^{-2}}-\frac{(1-p / q)\left(1-q^{3}\right)}{p / q}\right] \\
& =\frac{p^{3}}{q}\left[\frac{(q)_{3} q^{3}}{p^{3}} \frac{1}{(p)_{\infty}}-\frac{q^{2}(q-p)(q)_{3}}{p^{3}}-\frac{q(q-p)(q)_{3}}{p^{3}}-\frac{(q-p)\left(1-q^{3}\right)}{p}\right] \\
& =\frac{(q)_{3} q^{2}}{(p)_{\infty}}-q(q-p)(q)_{3}-(q-p)(q)_{3}-\frac{p^{2}(q-p)\left(1-q^{3}\right)}{q}
\end{aligned}
$$

Hence

$$
\begin{aligned}
\mathbb{E}\left(I_{2}\right) & =\frac{1}{p}\left[\frac{1}{(p)_{\infty}}-1\right]\left[2 p-1+\frac{q}{(p)_{\infty}}\right] \\
& +\frac{1}{p} \sum_{j \geq 0} \frac{p q^{j}}{1-p q^{j}}\left[\frac{q}{(p)_{j+1}}-\frac{q}{(p)_{\infty}}-\frac{p(1+q) q^{j}}{q(p)_{j+1}}\right]-\frac{1}{p} \sum_{j \geq 0} \frac{j p^{2} q^{2 j}}{(p)_{j+1}} \\
& +\frac{1}{p}\left[\frac{(q)_{3} q^{2}}{(p)_{\infty}}-q(q-p)(q)_{3}-(q-p)(q)_{3}-\frac{p^{2}(q-p)\left(1-q^{3}\right)}{q}\right]
\end{aligned}
$$

We still need (we use $S_{2}$ )

$$
\begin{aligned}
\sum_{j \geq 0} \frac{j q^{2 j}}{(p)_{j+1}} & =\sum_{h \geq 0} \sum_{j \geq h} \frac{q^{2 j}}{(p)_{j+1}} \\
& =\sum_{h \geq 0} \frac{q^{2 h}}{(p)_{h+1}} \sum_{j \geq 0} \frac{q^{2 j}}{\left(p q^{h+1}\right)_{j}} \\
& =\sum_{h \geq 0} \frac{q^{2 h}}{(p)_{h+1}}\left[-\frac{1}{p q^{2 h}}-\frac{1}{p q^{h-1}}+1+\frac{1}{p q^{2 h}\left(p q^{h+1}\right)_{\infty}}\right] \\
& =-\frac{1}{p} \sum_{h \geq 0} \frac{1}{(p)_{h+1}}\left[1-\frac{1}{\left(p q^{h+1}\right)_{\infty}}\right]-\frac{q}{p} \sum_{h \geq 0} \frac{q^{h}}{(p)_{h+1}}+\sum_{h \geq 0} \frac{q^{2 h}}{(p)_{h+1}} \\
& =\frac{1}{p q} \sum_{h \geq 0}\left[\frac{1}{(p q)_{\infty}}-\frac{1}{(p q)_{h}}\right]-\frac{1}{p}\left[\frac{1}{p(p q)_{\infty}}+1-\frac{1}{p}\right]+\frac{1}{q}\left[\frac{1}{p(p q)_{\infty}}-\frac{1}{p}-\frac{q}{p}+1\right] \\
& =\frac{1}{p q} \frac{1}{(p q)_{\infty}} \sum_{k \geq 1} \frac{\left[1-(p)_{k}\right] q^{k}}{1-q^{k}}-\frac{1}{p}\left[\frac{1}{p(p q)_{\infty}}+1-\frac{1}{p}\right]+\frac{1}{q}\left[\frac{1}{p(p q)_{\infty}}-\frac{1}{p}-\frac{q}{p}+1\right]
\end{aligned}
$$

hence

$$
\begin{aligned}
\mathbb{E}\left(I_{2}\right) & =\frac{1}{p}\left[\frac{1}{(p)_{\infty}}-1\right]\left[2 p-1+\frac{q}{(p)_{\infty}}\right] \\
& +\frac{1}{p} \sum_{j \geq 0} \frac{p q^{j}}{1-p q^{j}}\left[\frac{q}{(p)_{j+1}}-\frac{q}{(p)_{\infty}}-\frac{p(1+q) q^{j}}{q(p)_{j+1}}\right] \\
& -\frac{1}{p^{2} q} \frac{1}{(p q)_{\infty}} \sum_{k \geq 1} \frac{\left[1-(p)_{k}\right] q^{k}}{1-q^{k}}+\frac{1}{p^{2}}\left[\frac{1}{p(p q)_{\infty}}+1-\frac{1}{p}\right]-\frac{1}{p q}\left[\frac{1}{p(p q)_{\infty}}-\frac{1}{p}-\frac{q}{p}+1\right] \\
& +\frac{1}{p}\left[\frac{(q)_{3} q^{2}}{(p)_{\infty}}-q(q-p)(q)_{3}-(q-p)(q)_{3}-\frac{p^{2}(q-p)\left(1-q^{3}\right)}{q}\right] .
\end{aligned}
$$

Although this computation was already quite involved, we could get all cross-moments from (6.3), but only with considerable effort.

## 9 Markov Chains and Sojourn times

In this section, we obtain the asymptotic distribution of the number of descents initial values in some interval.

### 9.1 General asymptotic distribution

Let $X_{i}, i=1, \ldots, m$ be an ergodic Markov chain (MC) and $A$ be a subset of states. Assume that the MC is stationary and set $x_{i}:=\llbracket X_{i} \in A \rrbracket$, with $M:=\mathbb{E}\left(x_{i}\right)$. The number $D$ of times the MC is in $A$ on $1, \ldots, m$ is such that

$$
\begin{aligned}
\mathbb{E}(D) & =m M \\
\mathbb{V}(D) & =\mathbb{E}\left[\sum_{i=1}^{m}\left(x_{i}-M\right)^{2}+2 \sum_{i=1}^{m-1} \sum_{j=i+1}^{m}\left(x_{i}-M\right)\left(x_{j}-M\right)\right] .
\end{aligned}
$$

If

$$
\mathbb{E}\left[\left(x_{i}-M\right)\left(x_{j}-M\right)\right]=0, \quad j \geq i+2
$$

then, setting $B:=\mathbb{E}\left(x_{i} x_{i+1}\right)$, we obtain

$$
\mathbb{V}(D)=m M(1-M)+2(m-1)\left(B-M^{2}\right)
$$

But we have a central limit theorem for MC (again, see (3)). This gives

$$
\frac{D-\mathbb{E}(D)}{\sqrt{\mathbb{V}(D)}} \sim \mathcal{N}(0,1), \quad m \rightarrow \infty
$$

### 9.2 Number of descents values in some interval

Here, the states of the MC are couples $Y_{t} Y_{t+1}$ with geometric distribution. Also the MC starts with the stationary distribution $p q^{i-1} p q^{j-1}$, hence $m=n-1$.

If we are interested in the asymptotic (gaussian) distribution of the number of descents initial values in some interval $[\tau, \tau+\Delta]$, we compute

$$
\begin{aligned}
M & =\sum_{i=\tau}^{\tau+\Delta} p q^{i-1}\left(1-q^{i-1}\right) \\
B & =\sum_{i=\tau+1}^{\tau+\Delta} p q^{i-1} \sum_{j=\tau}^{i-1} p q^{j-1}\left(1-q^{j-1}\right)
\end{aligned}
$$

Of course, an explicit formula could be written for $M$ resp. $B$.

## 10 Permutations

Starting from $n$ geom $(p)$ RVs, as $q \rightarrow 1$, we can derive the asymptotic properties of first and second descents, in a large permutation. We will consider large size $(n \rightarrow \infty)$ permutations of $\{1, \ldots, n\}$, or $n$-permutations for short. It is well known that all rank statistics of an $n$-permutation can be derived from the corresponding ones of a sequence of $n$ geom $(p) \mathrm{RVs}$ as $q \rightarrow 1$. But, of course, it is not possible to deduce the moments of the beginning (initial height) of the first descent of a $n$-permutation from the corresponding moments of the geometric variables statistic, as $q \rightarrow 1$. However, the asymptotic distribution of this RV can be derived as follows. Let $q=1-\varepsilon$. This gives an asymptotic distribution function for each geom $\operatorname{RV} K(\varepsilon \rightarrow 0)$

$$
1-q^{i} \sim F(i)=1-e^{-i \varepsilon}
$$

Set $U=F(K) ; U$ is asymptotically distributed as a uniform[0,1] RV. Let us consider $n$ geom $(p)$ RVs $K_{1}, \ldots, K_{n}$. If we scale the corresponding $U$ variables with $n$ (i.e. multiply by $n$ ), take the integer part, we have asymptotically a permutation on $n$, for large $n$. More precisely the rank of $U_{i}$ is the value of the $i$ th element of the permutation. This gives

$$
u=1-e^{-i \varepsilon}, \quad d u=\varepsilon(1-u) d i, \quad i=-\ln (1-u) / \varepsilon
$$

Set $g(u):=f_{1}(i)$ as given by (7.1). Using $n$ geometric RVs instead of an infinite sequence of ones intoduces only an exponentially small error.

For instance, this leads to

$$
\mathbb{E}\left(I_{1}\right) \sim n \int_{0}^{1} g(u) u \frac{d u}{\varepsilon(1-u)}, \quad \varepsilon \rightarrow 0
$$

We have by simple algebra,

$$
g(u) \sim \varepsilon(1-u) u+e^{u} u\left(1-e^{-u}\right) \varepsilon(1-u)
$$

and the asymptotic density of $I_{1} / n$ is given by (we multiply by $d i=d u /(\varepsilon(1-u))$ )

$$
h_{1}(u)=u\left[1+e^{u}\left(1-e^{-u}\right)\right]=u e^{u}=u \sum_{j=0}^{\infty} \frac{u^{j}}{j!} .
$$

But this has a clear direct probabilistic interpretation: if $U_{j+1}=u$, the probability

$$
\mathbb{P}\left[U_{1} \leq U_{2} \leq \cdots \leq U_{j} \leq U_{j+1} \geq U_{j+2}\right]=u \frac{u^{j}}{j!}
$$

Again, as $n$ is large, we can use an infinite summation on $j$, with exponentially small error.
Note that

$$
\int_{0}^{1} h_{1}(u) d u=1
$$

as expected, and

$$
\mathbb{E}\left(I_{1}\right) \sim n \int_{0}^{1} h_{1}(u) u d u=n(e-2):=n E_{1,0}, \quad n \rightarrow \infty
$$

which conforms to Theorem 9 of (5). All moments can be derived. For instance

$$
\begin{aligned}
& \mathbb{E}\left(I_{1}^{2}\right) \sim n^{2} \int_{0}^{1} h_{1}(u) u^{2} d u=n^{2}(6-2 e):=n^{2} E_{2,0} \\
& \mathbb{E}\left(I_{1}^{3}\right) \sim n^{3} E_{3,0}, \text { with } E_{3,0}=9 e-24
\end{aligned}
$$

A general expression can be derived as follows. We have (we drop the second index)

$$
E_{k}=e-(k+1) E_{k-1}, \quad E_{0}=1
$$

This first order recursion can be solved by iteration:

$$
\begin{equation*}
E_{k}=(k+1)!(-1)^{k}\left[e \sum_{i=1}^{k} \frac{(-1)^{i}}{(i+1)!}+1\right] \tag{10.1}
\end{equation*}
$$

By convention, $E_{-1}=\int_{0}^{1} e^{u} d u=e-1$.
We could derive similar expressions for $J_{2}$.
The errors terms, as $n \rightarrow \infty$ can be derived as follows. The rank $R$ of $U_{j+1}$ is such that $R-(j+2)$ is Binomial $(n-(j+2), u)$, if $U_{j+1}$ has value $u$ and corresponds to the first descent. Indeed, $j$ values are already below $u$ (on the left) and one on the right. If $d$ values among the $n-(j+2)$ remaining ones are below $u$, then $U_{j+1}$ possesses the rank $d+j+2$. So the moments of $R / n$ are given by the characteristic function

$$
F:=\left[1+u\left(e^{\mathbf{i} \theta / n}-1\right)\right]^{n-(j+2)} e^{\mathbf{i} \theta(j+2) / n}
$$

from which we derive

$$
\mathbb{E}(R / n \mid j) \sim u+\frac{-u(u-1)(j+2)}{n}
$$

$$
\mathbb{E}\left(R^{2} / n^{2} \mid j\right) \sim u^{2}+\frac{-u(u-1)(5+2 j)}{n}
$$

We multiply by $u \frac{u^{j}}{j!}$, and sum on $j$; this gives

$$
\begin{aligned}
\sum_{j=0}^{\infty} u \frac{u^{j}}{j!} \mathbb{E}(R / n \mid j) & \sim h_{1}(u) u+\frac{-e^{u} u(u-1)(u+2)}{n} \\
\sum_{j=0}^{\infty} u \frac{u^{j}}{j!} \mathbb{E}\left(R^{2} / n^{2} \mid j\right) & \sim h_{1}(u) u^{2}+\frac{-e^{u} u^{2}(u-1)(2 u+5)}{n}
\end{aligned}
$$

Hence, integrating on $u \in[0 . .1]$, we finally obtain

$$
\begin{aligned}
\mathbb{E}\left(I_{1} / n\right) & \sim E_{1,0}+\frac{e-2}{n} \\
\mathbb{E}\left(I_{1}^{2} / n^{2}\right) & \sim E_{2,0}+\frac{20-7 e}{n}
\end{aligned}
$$

The first expression fits with Theorem 9 in (5); the second moment wasn't computed in this paper.
The joint moments of the first and second descent $I_{1}, I_{2}$ are asymptotically computed as follows. First, from (6.3),

$$
\begin{aligned}
g_{1}\left(u_{2}, u_{1}\right) & =\llbracket u_{2}<u_{1} \rrbracket \varepsilon\left(1-u_{2}\right) \frac{u_{2}}{u_{1}}+\int_{0}^{\min \left(u_{1}, u_{2}\right)} \frac{d v_{1}}{\varepsilon\left(1-v_{1}\right)} \varepsilon \frac{1-v_{1}}{u_{1}} e^{u_{2}-v_{1}} \varepsilon u_{2}\left(1-u_{2}\right) \\
& =\llbracket u_{2}<u_{1} \rrbracket \varepsilon\left(1-u_{2}\right) \frac{u_{2}}{u_{1}}+e^{u_{2}} \varepsilon u_{2} \frac{1-u_{2}}{u_{1}}\left(1-e^{-\min \left(u_{1}, u_{2}\right)}\right) .
\end{aligned}
$$

This leads to the Markov kernel

$$
\begin{equation*}
h\left(u_{2}, u_{1}\right)=\frac{g_{1}\left(u_{2}, u_{1}\right)}{\varepsilon\left(1-u_{2}\right)}=\llbracket u_{2}<u_{1} \rrbracket \frac{u_{2}}{u_{1}}+e^{u_{2}} u_{2} \frac{1}{u_{1}}\left(1-e^{-\min \left(u_{1}, u_{2}\right)}\right) \tag{10.2}
\end{equation*}
$$

This can also be derived from the uniform $[0,1]$ random variables properties.
The joint moment is asymptotically given given by $n^{2} E_{k, l}$, with

$$
\begin{aligned}
E_{k, l} & =\int_{0}^{1} \int_{0}^{1} h_{1}\left(u_{1}\right) h\left(u_{2}, u_{1}\right) u_{1}^{k} u_{2}^{l} d u_{1} d u_{2} \\
& =\int_{0}^{1} u_{1} e^{u_{1}} \int_{0}^{1} u_{1}^{k} u_{2}^{l} \frac{1}{u_{1}}\left[\llbracket u_{2}<u_{1} \rrbracket u_{2}+e^{u_{2}} u_{2}\left(1-e^{-\min \left(u_{1}, u_{2}\right)}\right)\right] d u_{1} d u_{2}
\end{aligned}
$$

The first values of $E_{k, l}$ are given in Table 1.
The asymptotic density of $I_{2} / n$ is given by

$$
h_{2}(u)=\int_{0}^{1} h_{1}\left(u_{1}\right) h\left(u, u_{1}\right) d u_{1}=u\left[e^{1+u}-u e^{u}-e^{u}\right]=u e^{u}[-1-u+e]
$$

and similarly for $h_{3}(u)$. This gives eventually

$$
h_{3}(u)=\int_{0}^{1} h_{2}\left(u_{2}\right) h\left(u_{3}, u_{2}\right) d u_{2}=u e^{u}\left[\left(u+u^{2} / 2\right)+e(-u-2)+e^{2}\right] .
$$

| $k>l$ | 0 | 1 | 2 |
| :---: | :---: | :---: | :---: |
| 0 | 1 | $e^{2}-e-4$ | $-2 e^{2}-e+18$ |
| 1 | $e-2$ | $-7 e / 2+10$ | $20 e-54$ |
| 2 | $6-2 e$ | $e^{2}+32 e / 3-36$ | $-2 e^{2}-235 e / 3+228$ |

Tab. 1: $E_{k, l}$


Fig. 4: $h(u), h_{1}(u), h_{2}(u)$

The convergence to the stationary distribution $h(u)=2 u$ is very fast. $h(u)$ can be derived from (6.4), or directly by considering two successive RVs $U_{k}$. Figure 4 gives $h(u), h_{1}(u), h_{2}(u)$

The difference $h_{3}(u)-h(u)$ is given in Figure 5.
A general expression for $h_{i}(u)$ can be computed as follows: we observe that $h_{i}(u)$ is of the form

$$
h_{i}(u)=u e^{u} \sum_{k=0}^{i-1} e^{k} P_{i, k}(u)
$$

where

$$
P_{i, k}(u)=\sum_{l=0}^{i-k-1} P(i, k, l) u^{l}
$$

But, setting $E_{l}:=D_{l, 0}+D_{l, 1} e$, from (10.2) and (10.1),

$$
h_{i}(u)=\int_{0}^{1} h_{i-1}(u) h(u, v) d v
$$



Fig. 5: $h_{3}(u)-h(u)$

$$
\begin{aligned}
& =\int_{0}^{1} h_{i-1}(v) \frac{1}{v}\left[\llbracket u<v \rrbracket u+e^{u} u\left(1-e^{-\min (v, u)}\right)\right] d v \\
& =\int_{0}^{u} h_{i-1}(v) \frac{1}{v} e^{u} u\left(1-e^{-v}\right) d v+\int_{u}^{1} h_{i-1}(v) \frac{1}{v} e^{u} u d v \\
& =\int_{0}^{1} h_{i-1}(v) \frac{1}{v} e^{u} u d v-\int_{0}^{u} h_{i-1}(v) \frac{1}{v} e^{u} u e^{-v} d v \\
& =e^{u} u\left[\sum_{k=0}^{i-2} e^{k} \int_{0}^{1} P_{i-1, k}(v) e^{v} d v-\sum_{k=0}^{i-2} e^{k} \int_{0}^{u} P_{i-1, k}(v) d v\right] \\
& =e^{u} u\left[\sum_{k=0}^{i-2} e^{k} \sum_{l=0}^{i-k-2} P_{i-1, k, l}\left[D_{l-1,0}+D_{l-1,1} e\right]-\sum_{k=0}^{i-2} e^{k} \sum_{l=0}^{i-k-2} P_{i-1, k, l} \frac{u^{l+1}}{l+1}\right] .
\end{aligned}
$$

This leads to the following recurrence

$$
\begin{aligned}
P_{i, i-1}(u) & =1 \\
P_{i, 0,0} & =\sum_{l=0}^{i-2} P_{i-1,0, l} D_{l-1,0}, \\
P_{i, k, 0} & =\sum_{l=0}^{i-k-2} P_{i-1, k, l} D_{l-1,0}+\sum_{l=0}^{i-k-1} P_{i-1, k-1, l} D_{l-1,1}, \quad k=1, \ldots, i-2, \\
P_{i, k, l} & =-P_{i-1, k, l-1} / l, \quad k=0, \ldots, i-2, l=1, \ldots, i-k-1
\end{aligned}
$$

Finally, asymptotically, given $I_{k}, J_{k}$ is uniform $\left[1 \ldots I_{k}-1\right]$.

The number of descents values of $I / n$ in some interval $[\tau, \tau+\Delta]$ is now related to

$$
\begin{aligned}
M & =\int_{\tau}^{\tau+\Delta} u d u \\
B & =\int_{\tau}^{\tau+\Delta} d u_{1} \int_{\tau}^{u_{1}} u_{2} d u_{2} .
\end{aligned}
$$

## 11 Conclusion

In this paper, we have made a combinatorial and probabilistic study of initial and end heights of first, second, ... descents in samples of geometrically distributed random variables and in permutations. Several other (similar) models can be analyzed with our tools, let us mention: the weak model and/or ascents with the probabilistic approach, $k$-descents, size $d$ or more descents, (see (2)), last descents, (see (4)), $k$-ascents with a combinatorial approach (see (2)), etc.

We leave these topics to future work (or to the interested reader), in order to keep the length of the paper within reasonable limits.

## Appendix

## A Some combinatorial identities

We will use intensively Heine's formula:

$$
\sum_{m \geq 0} \frac{(a)_{m}(b)_{m} t^{m}}{(q)_{m}(c)_{m}}=\frac{(b)_{\infty}(a t)_{\infty}}{(c)_{\infty}(t)_{\infty}} \sum_{m \geq 0} \frac{(c / b)_{m}(t)_{m} b^{m}}{(q)_{m}(a t)_{m}}
$$

Several $q$-combinatorial relations are deduced from Heine.

$$
\begin{align*}
p z \sum_{m \geq 1} \frac{q^{2 m}}{(p z q)_{m}} & =\frac{1}{z}\left[\frac{1}{(p q z)_{\infty}}-1\right]-q . \text { This is Thm. } 1 \text { in (5). }  \tag{A.1}\\
\sum_{m \geq 0}(p z)_{m} q^{m} & =\frac{1}{p z}\left[1-(p z)_{\infty}\right] . \text { This is (5) in (5). }  \tag{A.2}\\
p z \sum_{m \geq 1} \frac{q^{m}}{(p z)_{m+1}} & =\frac{1}{1-p z}\left[\frac{1}{(p q z)_{\infty}}-1\right] . \text { This is implicitly used in (5). }  \tag{A.3}\\
& \text { Heine with } t=q, a=0, b=q, c=p z q, \text { and (A.2). } \\
\sum_{m \geq 1} \frac{p q^{m}\left(1-q^{m}\right)}{(p)_{m+1}} & =1 . \text { Application of (A.3), (A.1). }  \tag{A.4}\\
\sum_{m \geq 0}(p z)_{m} q^{2 m} & =\frac{1}{(p z)^{2}}\left[1-(p z)_{\infty}-\frac{1-p z}{q}+\frac{(p z)_{\infty}}{q}\right] . \text { This is (6) in (5). } \tag{A.5}
\end{align*}
$$

$$
\begin{align*}
\sum_{k \geq 1} \frac{p^{k}}{(q)_{k}} & =-1+\frac{1}{(p)_{\infty}} . \text { Heine with } t=p, a=0, b=q, c=q .  \tag{A.6}\\
\sum_{k \geq 0} \frac{t^{k}}{(p q)_{k}} & =\frac{(q)_{\infty}}{(p q)_{\infty}(t)_{\infty}} \sum_{m \geq 1} \frac{(p)_{m}(t)_{m} q^{m}}{(q)_{m}} . \text { Heine with } a=0, b=q, c=p q .  \tag{A.7}\\
(p)_{j-1} \sum_{i=j}^{\infty} \frac{p q^{i-1}\left(1-q^{i-1}\right)}{(p)_{i}} & =q^{j-1} . \text { Application of (A.1), (A.3). }  \tag{A.8}\\
\sum_{0 \leq j<J} p q^{j}(p)_{j} & =p \sum_{j \geq 0} q^{j}(p)_{j}-p \sum_{j \geq J} q^{j}(p)_{j} \\
& =p \frac{1}{p}\left[1-(p)_{\infty}\right]-p q^{J}(p)_{J} \frac{1}{p q^{J}}\left[1-\left(p q^{J}\right)_{\infty}\right] \\
& =1-(p)_{\infty}-(p)_{J}\left[1-\left(p q^{J}\right)_{\infty}\right] \\
& =1-(p)_{J} . \mathrm{By}(\mathrm{~A} .2) .  \tag{A.9}\\
\sum_{i_{1}>i_{0}} \frac{p q^{i_{1}-1}}{(p)_{i_{1}}} & =\frac{1}{(p)_{i_{0}-1}\left(1-p q^{i_{0}-1}\right)}\left[\frac{1}{\left(p q^{i_{0}}\right)_{\infty}}-1\right] \\
& =\frac{1}{(p)_{i_{0}}}\left[\frac{1}{\left(p q^{i_{0}}\right)_{\infty}}-1\right] \\
& =\left[\frac{1}{(p)_{\infty}}-\frac{1}{(p)_{i_{0}}}\right] . \text { By (A.3). } \tag{A.10}
\end{align*}
$$

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