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# Generation of a Random Partition of a Finite Set by an Urn Model 

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A random partition of $\mathbb{N}_{n}=\{1, \ldots, n\}$ may be generated by putting the elements of $\mathbb{N}_{n}$ at randon into a stochastic number of cells. This representations is used to prove asymptotic results about the random partition for $n \rightarrow \infty$.

## 1. Introduction and Notations

A partition of $N_{n}=\{1, \ldots, n\}$ is an unordered family of disjoint proper subset of $\mathbb{N}_{n}$ with union $\mathbb{N}_{n}$. A generic element of the class $I_{n}$ of partitions of $\mathbb{N}_{n}$ will be denoted by $\pi$. We have $\left|I_{n}\right|=T_{n}$, where $T_{n}$ is the $n$th Bell or exponential number given by Dobinski's formula

$$
\begin{equation*}
T_{n}=e^{-1} \sum_{k=1}^{\infty} k^{n} / k!. \tag{1.1}
\end{equation*}
$$

The number of partitions of $\mathbb{N}_{n}$ consisting of $k$ subsets is $S(n, k)$, a Stirling number of the second kind. See Comtet [2]. We define $S(n, k)=0$ for $k>n$ and for $k \leqslant 0, n \geqslant 1$. The asymptotic behaviour as $n \rightarrow \infty$ of $T_{n}$, originally derived by Moser and Wyman [9], and the combinatorics of $\Pi_{n}$ were studied in $[3,4,12-15]$. We will need the relation

$$
\begin{equation*}
T_{n+1} / T_{n}=n \rho_{n}^{-1}+\frac{1}{2} \rho_{n}\left(1+\rho_{n}\right)^{-2}+O\left(n^{-1} \rho_{n}\right), \tag{1.2}
\end{equation*}
$$

where $\rho_{n}=\rho(n)$ and $\rho$ is the inverse of the function $x e^{x}$ on $[0, \infty)$, so that $\rho_{n} \sim \log n$ as $n \rightarrow \infty$. Manipulation of the $\rho_{n}$ is simplified by the relation

$$
\begin{equation*}
\rho^{\prime}(t)=t^{-1} \rho(t) /(1+\rho(t)) . \tag{1.3}
\end{equation*}
$$

The combinatorics of $\Pi_{n}$ may be described in terms of probabiiity theory by "choosing an element of $\Pi_{n}$ at random," i.e., defining a probability
measure $P_{n}$ on $\Pi_{n}$, or rather on the power set of $\Pi_{n}$, such that $P_{n}(\{\pi\})=T_{n}^{-1}, \pi \in \Pi_{n}$. The number of partitions having property $A$ then is $T_{n} P_{n}(A)$, where $P_{n}(A)$ is the probability that the random partition has property $A$. The heuristic value of this description lies in the exploitation of concepts and theorems of probability theory, such as independence and the central limit theorem, as in [1, 4]. More may be done by embedding ( $\Pi_{n}, P_{n}$ ) into a larger probability space $(\Omega, \mathscr{S}, P)$, where $P$ is a probability measure on the $\sigma$-algebra $\mathscr{S}$ of subsets of $\Omega$, and $\pi$ is a stochastic variable on $\Omega$. The most obvious method would be the urn model with a fixed number $v$ of urns. Put the elements of $N_{n}$ at random into the urns, i.e., place them independently with probability $v^{-1}$ for each of the urns. The nonempty urns then define a stochastic partition of $\mathbb{N}_{n}$. This, however, is not a random partition: the resulting probability distribution on $\Pi_{n}$ is not the uniform distribution $P_{n}$, see $[5,7]$.

We may obtain $P_{n}$ by taking a stochastic number $U$ of urns with

$$
\begin{equation*}
P(U=u)=q_{n u}=T_{n}^{-1} e^{-1} u^{n} / u!, \quad u=1,2, \ldots \tag{1.4}
\end{equation*}
$$

Note that (1.4) defines a probability distribution by (1.1). Under the condition $U=u$ the elements of $\mathbb{N}_{n}$ are put at random into the $u$ urns. The contents of the nonempty urns then form a stochastic partition of $\mathbb{N}_{n}$. We have

Theorem 1. This partition is a random partition of $\mathbb{N}_{n}$, i.e., the probability that the urn procedure generates a prescribed $\pi \in \Pi_{n}$ is $T_{n}^{-1}$.

Proof. If $\pi$ consists of $k$ subsets, the conditional probability $P(\pi \mid u)$ of obtaining $\pi$ given $U=u$ is $u^{(k)} u^{-n}$ for $u \geqslant k$ and zero for $u<k$, where

$$
\begin{equation*}
m^{(0)}=1, \quad m^{(i)}=m(m-1) \cdots(m-i+1), \quad m, i \in \mathbb{N} \tag{1.5}
\end{equation*}
$$

So with (1.4) and (1.1),

$$
P(\pi)=\sum_{u=k}^{\infty} q_{n u} u^{(k)} u^{-n}=T_{n}^{-1} \sum_{u=k}^{\infty} e^{-1} /(u-k)!=T_{n}^{-1} .
$$

A formal definition of the above probability measure $P$ may be given on a countable product space $\Omega$, see Neveu [10, Chap. III].

Theorem 1 implies that any theorem about $P_{n}$ on $\Pi_{n}$ may be derived by considering the random partition to be generated by the urn process. This method, with a different urn model, was applied by Kolchin [6] to the cycles of a random permutation of $\mathbb{N}_{n}$.

The urn model will be applied to the following problems. In the random partition $\pi$ let $A_{i}$ be the element of $\pi$ that contains $i$ and let

$$
\begin{equation*}
W_{i}=\left|A_{i}\right|-1, \quad i=1, \ldots, n \tag{1.6}
\end{equation*}
$$

Since a partition defines an equivalence relation on $\mathbb{N}_{n}$, the stochastic variable $W_{i}$ also may be defined as the number of elements of $\mathbb{N}_{n}$ that are equivalent to $i$. In Section 2 we will study the properties of the probability distribution (1.4) and show that the $W_{i}$ are asymptotically independent with Poisson distribution.

Let $\pi=\left\{B_{1}, \ldots, B_{k}\right\} \in \Pi_{n}$. Then, if $r \leqslant n$, the nonempty sets $B_{j} \cap \mathbb{N}_{r}$ form a partition of $\mathbb{N}_{r}$, called the relative partition of $\mathbb{N}_{r}$ induced by $\pi$, and denoted $\pi \cap \mathbb{N}_{r}$. If $\pi$ is random, $\pi \cap \mathbb{N}_{r}$ in general is not, i.e., its probability distribution is not $P_{r}$. In Section 3 we prove that asymptotically as $n \rightarrow \infty$ and $r / n=O\left(\rho_{n}^{-1}\right)$, the relative partition may be considered to be generated by an urn model with a nonstochastic number of urns.

These results will be formulated in terms of $L_{1}$, the absolute variation distance between probability measures. Let $m_{1}, m_{2}$ be finite measures on $B \subset \mathbb{Z}^{k}$ with

$$
\begin{equation*}
m_{i}(A)=\sum_{j \in A} p_{i}(j), \quad A \subset B, \quad i=1,2, \tag{1.7}
\end{equation*}
$$

where, as later on, we write $j$ for $\left(j_{1}, \ldots, j_{k}\right)$. The $L_{1}$ distance between $m_{1}$ and $m_{2}$ is

$$
\begin{equation*}
\left\|m_{1}-m_{2}\right\|=\sum_{j \in B}\left|p_{1}(j)-p_{2}(j)\right| . \tag{1.8}
\end{equation*}
$$

For $\psi: B \rightarrow D \subset \mathbb{Z}^{h}$ we define the measures $m_{i} \psi^{-1}$ on $D$ by $m_{i} \psi^{-1}(G)=$ $m_{i}\left(\psi^{-1}(G)\right)$. If $m_{i}$ is a probability measure, $m_{i} \psi^{-1}$ is the probability distribution of $\psi$ under $m_{i}$. We have the following lemma, easily proved:

Lemma 1.

$$
\left\|m_{1} \psi^{-1}-m_{2} \psi^{-1}\right\| \leqslant\left\|m_{1}-m_{2}\right\|,
$$

with equality if $p_{1}(j) / p_{2}(j)=f \circ \psi(j)$ for $p_{1}(j)+p_{2}(j)>0$, where we define $a / 0=\infty, a>0$.

## 2. The Urn Model and the $W_{i}$

For the urn model described in Section 1 we have

Theorem 2. The random partition of $\mathbb{N}_{n}$ generated by the urn model and the number $M_{0}$ of empty urns are independent. The distribution of $M_{0}$ is Poisson with parameter 1 .

Proof. In order that $M_{0}=m$ and that a prescribed partition $\pi$ consisting of $k$ sets is generated, we must have $U=m+k$. So

$$
\begin{aligned}
P\left(\pi, M_{0}=m\right) & =P(U=m+k) P\left(\pi, M_{0}=m \mid U=m+k\right) \\
& =T_{n}^{-1} e^{-1}(m+k)^{n}(m+k)^{(k)}(m+k)^{-n} /(m+k)! \\
& =T_{n}^{-1} e^{-1} / m!=P(\pi) e^{-1} / m!
\end{aligned}
$$

For the generating function and the moments of $U$ we find from (1.4) and (1.1)

$$
\begin{align*}
& E z^{U}=\exp (z-1) T_{n}^{-1} \sum_{u=1}^{\infty} z^{u} e^{-z} u^{n} / u!  \tag{2.1}\\
& E U^{h}=T_{n+h} / T_{n} \tag{2.2}
\end{align*}
$$

From (2.2) and (1.2) with (1.3)

$$
\begin{align*}
& a_{n}-E U-n \rho_{n}^{-1}+\frac{1}{2} \rho_{n}\left(1+\rho_{n}\right)^{-2}+O\left(n^{-1} \rho_{n}\right)  \tag{2.3}\\
& \sigma_{n}^{2}=\sigma^{2}(U)=n \rho_{n}^{-1}\left(1+\rho_{n}\right)^{-1}+O(1) \tag{2.4}
\end{align*}
$$

Remark. Let $X$ be the number of subsets of $N_{n}$ in the random partition. Since $X$ and $M_{0}$ are independent by Theorem 2 and $U=X+M_{0}$, relation (2.1) corresponds to the convolution property of generating functions and we may use (2.2) to find lower order moments of $X$.

Let $W_{i}$ be defined by (1.6). By conditioning with respect to $U$ and the urn into which $i$ lands, we see that the distribution of $W_{i}$ is weighted binomially. For $k=0, \ldots, n-1$,

$$
\begin{equation*}
P_{n}\left(W_{i}=k\right)=\sum_{n=1}^{\infty} q_{n u}\binom{n-1}{k} u^{-k}\left(1-u^{-1}\right)^{n-1-k} \tag{2.5}
\end{equation*}
$$

From (2.5) with (1.1)-(1.3)

$$
\begin{align*}
b_{n}= & E W_{i}=(n-1) T_{n-1} / T_{n} \\
= & \rho_{n}\left\{1-(2 n)^{-1}-(2 n)^{-1}\left(1+\rho_{n}\right)^{-2}+O\left(n^{-2} \rho_{n}^{2}\right)\right\},  \tag{2.6}\\
\sigma^{2}\left(W_{i}\right)= & (n-1)^{2}\left(T_{n-2} T_{n}^{-1}-T_{n-1}^{2} T_{n}^{-2}\right) \\
& +(n-1) T_{n}^{-1}\left(T_{n-1}-T_{n-2}\right)=b_{n}\left(1+b_{n-1}-b_{n}\right) \\
= & \rho_{n}\left\{1-(2 n)^{-1}\left(3 \rho_{n}^{2}+4 \rho_{n}+2\right)\left(1+\rho_{n}\right)^{-2}+O\left(n^{-2} \rho_{n}^{2}\right)\right\} . \tag{2.7}
\end{align*}
$$

Now let $Q=Q_{n n}$ be the joint probability distribution of $W_{1}, \ldots, W_{n}$ and $R=R_{n h}$ the product of $h$ Poisson probability distributions, each with parameter $\rho_{n}$. Then we have

Theorem 3. With $L_{1}$ distance defined by (1.8) and $a_{n}, \sigma_{n}$ by (2.3) and (2.4),

$$
\begin{equation*}
\|Q-R\| \leqslant h T_{n}^{-1} T_{n-1}\left(h+2+2\left\lfloor\rho_{n}^{2} \sigma_{n-1}^{2}+\left.\left(\rho_{n} a_{n-1}-n+h\right)^{2}\right|^{1 / 2}\right\rangle .\right. \tag{2.8}
\end{equation*}
$$

Remark. If $h=o\left(n^{1 / 2} \rho_{n}^{-1}\right)$ as $n \rightarrow \infty$, we have by (1.2), (2.3), (2.4)

$$
\begin{equation*}
\|Q-R\| \sim 2 n^{-1 / 2} \rho_{n} h \rightarrow 0, \quad n \rightarrow \infty . \tag{2.9}
\end{equation*}
$$

From the central limit theorem for the Poisson distribution it follows that $W_{1}, \ldots, W_{k}$ are asymptotically normal and independent.
Proof. We write $W=\left(W_{1}, \ldots, W_{h}\right)$ and $j=\left(j_{1}, \ldots, j_{h}\right)$. Let $D$ be the event that $1, \ldots, h$ are put into different urns. We have

$$
\begin{align*}
P(D) & =\sum_{u=h}^{\infty} q_{n u} u^{(h)} u^{-h} \\
& \geqslant \sum_{u=1}^{\infty} q_{n u}\left\{1-\frac{1}{2} h(h-1) u^{-1}\right\}=1-\frac{1}{2} h(h-1) T_{n-1} T_{n}^{-1} \tag{2.10}
\end{align*}
$$

By conditioning with respect to the urns into which $1, \ldots, h$ are put, we have

$$
\begin{equation*}
P(D, W=j)=\sum_{u=h}^{\infty} q_{n u} u^{(h)} u^{-h} a(n-h, u, j), \tag{2.11}
\end{equation*}
$$

where the $a(n-h, u, j)$ are probabilities of a multinomial distribution $A=$ $A(n-h, u)$ defined by

$$
\begin{equation*}
a(n-h, u, j)=\frac{(n-h)!}{j_{1}!\cdots j_{h}!(n-h-k)!} u^{-k}\left(1-h u^{-1}\right)^{n-h-k}, \tag{2.12}
\end{equation*}
$$

for $0 \leqslant k=j_{1}+\cdots+j_{h} \leqslant n-h$ and zero elsewhere. Then, with (2.10) and (2.11)

$$
\begin{align*}
\|Q-R\| & =\sum_{j}\left|P(D, W-j)+P\left(D^{c}, W=j\right)-R(j)\right| \\
& \leqslant \sum_{j}|P(D, W=j)-P(D) R(j)|+2 P\left(D^{c}\right) \\
& \leqslant 2 P\left(D^{c}\right)+\sum_{u=h}^{\infty} q_{n u} u^{(h)} u^{-h}\|A(n-h, u)-R\| . \tag{2.13}
\end{align*}
$$

Let $B=B(n-h, u)$ be the product of $h$ Poisson distributions with parameter $(n-h) u^{-1}$, i.e., with the same first moments as the multinomial distribution $A(n-h, u)$. We have

$$
\begin{equation*}
\|A-R\| \leqslant\|A-B\|+\|B-R\| . \tag{2.14}
\end{equation*}
$$

It is easily seen that $A(n-h, u)$ and $B(n-h, u)$ satisfy the equality condition of Lemma 1 with $\psi(j)=j_{1}+\cdots+j_{h}$, and that $A \psi^{-1}$ is the binomial distribution with parameters $n-h$ and $h u^{-1}$, whereas $B \psi^{-1}$ is the Poisson distribution with parameter $(n-h) h u^{-1}$. It was proved in [8,11] that $\left\|A \psi^{-1}-B \psi^{-1}\right\| \leqslant 3 h u^{-1}$. So

$$
\begin{equation*}
\|A(n-h, u)-B(n-h, u)\| \leqslant 3 h u^{-1} . \tag{2.15}
\end{equation*}
$$

Writing $f_{i}(j)=u_{i}^{k}\left(j_{1}!\cdots j_{h}!\right)^{-1}$ with $k=j_{1}+\cdots+j_{h}, i=1,2$, we have

$$
\begin{aligned}
&{\underset{j}{j}}\left|f_{1}(j) \exp \left(-h \alpha_{1}\right)-f_{2}(j) \exp \left(-h \alpha_{2}\right)\right| \\
& \leqslant \frac{\zeta_{j}}{} f_{1}(j)\left|\exp \left(-h \alpha_{1}\right)-\exp \left(-h \alpha_{2}\right)\right|+\frac{\Gamma}{j}\left|f_{1}(j)-f_{2}(j)\right| \exp \left(-h \alpha_{2}\right) \\
&=2-2 \exp \left(-h\left|\alpha_{1}-\alpha_{2}\right|\right) \\
& \leqslant 2 h\left|\alpha_{1}-\alpha_{2}\right|
\end{aligned}
$$

This gives, by definition of the Poisson distributions $R$ and $B(n-h, u)$,

$$
\begin{equation*}
\|R-B\| \leqslant 2 h\left|\rho_{n}-(n-h) u^{-1}\right| . \tag{2.16}
\end{equation*}
$$

From (2.13)-(2.16), with (1.1) and (1.4),

$$
\begin{align*}
\|Q-R\| & \leqslant 2 P\left(D^{c}\right)+3 h T_{n}^{-1} T_{n-1}+C_{n},  \tag{2.17}\\
C_{n} & =2 h \sum_{u=h}^{\infty} q_{n u} u^{(h)} u^{-h}\left|\rho_{n}-(n-h) u^{-1}\right| \\
& \leqslant 2 h \rho_{n} T_{n-1} T_{n}^{-1} \sum_{u=1}^{\infty} q_{n-1, u}\left|u-\rho_{n}^{-1}(n-h)\right| \\
& \leqslant 2 h \rho T_{n-1} T_{n}^{-1}\left\{\sum_{u=1}^{\infty} q_{n-1, u}\left(u-\rho_{n}^{-1}(n-h)\right)^{2}\right\}^{1 / 2} \\
& =2 h \rho_{n} T_{n-1} T_{n}^{-1}\left\{\sigma_{n-1}^{2}+\left(a_{n-1}-\rho_{n}^{-1}(n-h)\right)^{2}\right\}^{1 / 2}, \tag{2.18}
\end{align*}
$$

with $a_{n}$ and $\sigma_{n}$ given by (2.3) and (2.4). Inequality (2.8) now follows from (2.7), (2.10), and (2.18).

## 3. Relative Partitions

We need some results about the urn model with a fixed number $v$ of urns, see $[5,7]$. When $r$ objects are put at random into $v$ urns we have for $Y=Y_{r v}$, the number of occupied urns,

$$
\begin{align*}
p(r, v, k) & =P(Y=k)=S(r, k) v^{(k)} v^{-r}  \tag{3.1}\\
E Y & =v-v\left(1-v^{-1}\right)^{r}  \tag{3.2}\\
\sigma^{2}(Y) & =v\left(1-v^{-1}\right)^{r}+v(v-1)\left(1-2 v^{-1}\right)^{r}-v^{2}\left(1-v^{-1}\right)^{2 r} . \tag{3.3}
\end{align*}
$$

If $r \rightarrow \infty, v \rightarrow \infty$ so that $r / v \rightarrow \alpha \in(0, \infty)$, we have

$$
\begin{align*}
E Y & =v \theta(r / v)+O(1)  \tag{3.4}\\
\theta(x) & =1-e^{-x}  \tag{3.5}\\
\sigma^{2}(Y) & =v \eta(r / v)+O(1)  \tag{3.6}\\
\eta(x) & =e^{-2 x}\left(e^{x}-1-x\right) \tag{3.7}
\end{align*}
$$

As defined in Section 1 , a random partition of $\AA_{n}$ induces a relative partition of $\mathbb{N}_{r}$ that has a nonuniform probability distribution $G=G_{n r}$ on $\Pi_{r}$. We may consider $G_{n r}$ to be derived from the probability space ( $\Pi_{n}, P_{n}$ ), but also from the probability space $(\Omega, P)$ described under (1.4). It follows from (3.1) and (1.4) that the probability of $k$ subsets in the relative partition is

$$
\begin{equation*}
\sum_{u=k}^{\infty} q_{n u} S(r, k) u^{(k)} u^{-r}=S(r, k) T_{n}^{-1} \sum_{r=0}^{\infty} e^{-1}(v+k)^{n-r} / v! \tag{3.8}
\end{equation*}
$$

We will compare $G_{n r}$ with the probability measure $H_{n r}$ on $\Pi_{r}$ that corresponds to an urn model with a fixed number of urns. Put the elements of $\mathbb{N}_{r}$ at random into $c_{n}$ urns, where

$$
\begin{equation*}
c_{n}=\left[a_{n}\right], \quad a_{n}=E U \tag{3.9}
\end{equation*}
$$

see (1.4) and (2.3). The contents of the nonempty urns form a partition of $\mathbb{N}_{r}$ with probability distribution $H_{n r}$. The idea is that if $n \rightarrow \infty$, and $r / n \rightarrow 0$ sufficiently fast, the ratio $U / a_{n}$ is sufficiently close to 1 . We have

Theorem 4. If $n \rightarrow \infty$ and $r$ varies with $n$ so that $r=O\left(a_{n}\right)$,

$$
\begin{equation*}
\lim _{n \rightarrow \infty}\left\|G_{n r}-H_{n r}\right\|=0 \tag{3.10}
\end{equation*}
$$

Remark. This means that asymptotic results for relative partitions of a random partition of $\mathbb{N}_{n}$ may be derived from the urn model with a nonstochastic number of urns, e.g., Poisson and normal limits for distribution (3.8) as given in $[5,7]$.

Proof. First we take $r=\lambda c_{n}$, where $\lambda$ is a fixed positive integer. For a given partition $\pi \in \Pi_{r}$ consisting of $k$ subsets

$$
\begin{aligned}
& H_{n r}(\pi)=c_{n}^{(k)} c_{n}^{-r} \\
& G_{n r}(\pi)=\sum_{u=1}^{\infty} q_{n u} u^{(k)} u^{-r} .
\end{aligned}
$$

Note that $u^{(k)}=0, u \leqslant k$. From Lemma 1, (3.1) and (3.8),

$$
\begin{align*}
& \left\|H_{n r}-G_{n r}\right\|=\frac{V_{k=1}^{r}\left|p\left(r, c_{n}, k\right)-\sum_{u=1}^{\infty} q_{n u} p(r, u, k)\right|}{} \quad \leqslant \sum_{u=1}^{\infty} q_{n u} V(r, n, u), \\
& V=V(r, n, u)=\sum_{k=1}^{r} \mid p\left(r, c_{n}, k\right)-p(r, u, k) . \tag{3.11}
\end{align*}
$$

Put

$$
\begin{equation*}
D_{n}=\frac{\Sigma \prime}{u} q_{n u} V(r, n, u), \quad E_{n}=\frac{\Sigma^{\prime \prime}}{u} q_{n u} V(r, n, u) \tag{3.13}
\end{equation*}
$$

where $\Sigma^{\prime}$ and $\sum^{\prime \prime}$ mean summation over $\left|u-c_{n}\right| \geqslant K_{1} \sigma_{n}$ and $\left|u-c_{n}\right|<$ $K_{1} \sigma_{n}$, respectively, with $\sigma_{n}=\sigma(U)$ given by (2.4). Since $V \leqslant 2$, we have by (3.9) and Chebyshev's inequality

$$
\begin{equation*}
D_{n} \leqslant 2\left(K_{1}-1\right)^{-2} \tag{3.14}
\end{equation*}
$$

For $\left|u-c_{n}\right|<K_{1} \sigma_{n}$ we write

$$
\begin{gather*}
V(r, n, u)=\sum_{i=1}^{3} V_{i}(r, n, u),  \tag{3.15}\\
V_{1}=\frac{\_{k}}{}\left|p\left(r, c_{n}, k\right)-f(r, n, k)\right|,  \tag{3.16}\\
V_{2}=\grave{V}_{k} \mid f(r, n, k)-g(r, u, k),  \tag{3.17}\\
V_{3}=\grave{\}_{k}|g(r, u, k)-p(r, u, k)|, \tag{3.18}
\end{gather*}
$$

where we have, with $\beta=r / u$ and $\theta, \eta$ given by (3.5) and (3.7),

$$
\begin{align*}
& f=\left(2 \pi c_{n} \eta(\lambda)\right)^{-1 / 2} \exp \left\{-\left(k-c_{n} \theta(\lambda)\right)^{2} / 2 c_{n} \eta(\lambda)\right\}  \tag{3.19}\\
& g=(2 \pi u \eta(\beta))^{1 / 2} \exp \left\{-(k-u \theta(\beta))^{2} / 2 u \eta(\beta)\right\} \tag{3.20}
\end{align*}
$$

We have

$$
V_{1} \leqslant \Sigma^{(1)}\left|p\left(r, c_{n}, k\right)-f(r, n, k)\right|+\Sigma^{(2)}\left\{p\left(r, c_{n}, k\right)+f(r, n, k)\right\}
$$

where $\Sigma^{(1)}$ sums over $\left|k-c_{n} \theta(\lambda)\right| \leqslant K_{2} c_{n}^{1 / 2}$ and $\Sigma^{(2)}$ over $\left|k-c_{n} \theta(\lambda)\right|>$ $K_{2} c_{n}^{1 / 2}$. To the first term we apply the local central limit theorem [7, Chap. II.2, Theorem 2] for $p\left(r, c_{n}, k\right)$. Note that it applies to the number of empty urns. It shows that the first term is $O\left(c_{n}^{-1 / 2}\right)$. By Chebyshev's inequality, (3.1), and (3.4)-(3.7),

$$
\Sigma^{(2)} p\left(r, c_{n}, k\right) \leqslant \gamma_{1} K_{2}^{-2} \eta(\lambda) .
$$

By comparison with the integral over the corresponding normal density we find

$$
\Sigma^{(2)} f\left(r, c_{n}, k\right) \leqslant 2-2 \Phi\left(\eta^{-1 / 2}(\lambda)\left(K_{2}-c_{n}^{-1 / 2}\right)\right)
$$

where $\Phi$ is the standard normal distribution function. So

$$
\begin{equation*}
V_{1} \leqslant \gamma_{2}\left(K_{2}\right) c_{n}^{-1 / 2}+\gamma_{1} K_{2}^{-2} \eta(\lambda)+2-2 \Phi\left(\gamma_{3} K_{2}\right) . \tag{3.21}
\end{equation*}
$$

In (3.18) we split the domain of summation into $|k-u \theta(\beta)| \leqslant K_{2} u^{1 / 2}$ and $|k-u \theta(\beta)|>K_{2} u^{-1 / 2}$, with $\beta=r / u$. Since $\left|u-c_{n}\right|<K_{1} c_{n}$, it follows from (3.9), (2.3), (2.4) that $0<\gamma_{4}\left(K_{1}\right) \leqslant \beta \leqslant \gamma_{5}\left(K_{1}\right)<\infty$. We apply the local central limit theorem cited above to $p(r, u, k)$ and, since this theorem holds uniformly with respect to $\beta$ if $0<a \leqslant \beta \leqslant b<\infty$, we find in the same way as in (3.21)

$$
V_{3} \leqslant \gamma_{6}\left(K_{1}, K_{2}\right) u^{-1 / 2}+\gamma_{7} K_{2}^{-2} \eta(\beta)+2-2 \Phi\left(\eta^{-1 / 2}(\beta)\left(K_{2}-u^{-1 / 2}\right)\right)
$$

Again using the fact that $\left|u-c_{n}\right|<K_{1} \sigma_{n}$ and $r=\lambda c_{n}$ so that $\eta(\beta)$ is bounded, we see

$$
\begin{equation*}
V_{3} \leqslant \gamma_{7}\left(K_{1}, K_{2}\right) c_{n}^{-1 / 2}+\gamma_{8} K_{2}^{-2}+2-2 \Phi\left(\gamma_{9} K_{2}\right) . \tag{3.22}
\end{equation*}
$$

Comparing (3.17) with the $L_{1}$ distance between the corresponding normal densities and noting that $\left|u-c_{n}\right|<K_{1} \sigma_{n}$, we may show that

$$
\begin{equation*}
V_{2} \leqslant \gamma_{10}\left(K_{1}\right) \varepsilon_{n}, \quad \varepsilon_{n} \rightarrow 0, \quad n \rightarrow \infty . \tag{3.23}
\end{equation*}
$$

From (3.11), (3.13)-(3.15), (3.21)-(3.23), noting that the bounds in (3.21)-(3.23) do not depend on $u$, we see by first choosing $K_{1}$ and $K_{2}$ sufficiently large, that $\left\|G_{n r}-H_{n r}\right\|<\varepsilon$ for $n \geqslant n(\varepsilon)$.

Finally let us only assume $r=O\left(a_{n}\right)$. By (3.9) and (2.3) there is an integer $\lambda$ such that $r \leqslant R=\lambda c_{n} \leqslant n$ for $n \geqslant n_{1}$. Define the map $\psi: \Pi_{R} \rightarrow \Pi_{r}$ so that $\psi(\pi)=\pi \cap \mathbb{N}_{r}$. By Lemma 1

$$
\begin{equation*}
\left\|G_{n R} \psi^{-1}-H_{n R} \psi^{-1}\right\| \leqslant\left\|G_{n R}-H_{n R}\right\| . \tag{3.24}
\end{equation*}
$$

But $G_{n R}=P_{n} \chi^{-1}$, where $\chi: \Pi_{n} \rightarrow \Pi_{R}$ is defined by $\chi\left(\pi^{1}\right)=\pi^{1} \cap \mathbb{N}_{R}$. Since $\psi \circ \chi\left(\pi^{1}\right)=\pi^{1} \cap \mathbb{N}_{r}$, we have $G_{n r}=G_{n R} \psi^{-1}$. Since $H_{n R}$ is the probability distribution of the partion of $\mathbb{N}_{R}$ generated by throwing the elements of $\mathbb{N}_{R}$ at random into $c_{n}$ urns, $H_{n R} \psi^{-1}$ is the distribution of the partition of $\mathbb{N}_{r}$ generated by putting the elements of $\mathbb{N}_{r}$ at random into these urns, i.e., $H_{n k} \psi^{-1}=H_{n r}$. Relation (3.10) now follows from (3.24) and the first part of the proof.

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