# THE GENERALISED COUPON COLLECTOR PROBLEM

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#### Abstract

Coupons are collected one at a time from a population containing n distinct types of coupon. The process is repeated until all n coupons have been collected and the total number of draws, Y, from the population is recorded. It is assumed that the draws from the population are independent and identically distributed (draws with replacement) according to a probability distribution X with the probability that a type-i coupon is drawn being P(X = i). The special case where each type of coupon is equally likely to be drawn from the population is the classic coupon collector problem. We consider the asymptotic distribution Y (appropriately normalized) as the number of coupons  $n \to \infty$  under general assumptions upon the asymptotic distribution of X. The results are proved by studying the total number of coupons, W(t), not collected in t draws from the population and noting that  $P(Y \le t) = P(W(t) = 0)$ . Two normalizations of Y are considered, the choice of normalization depending upon whether or not a suitable Poisson limit exists for W(t). Finally, extensions to the K-coupon collector problem and the birthday problem are given.

Keywords: Coupon collector problem; Poisson convergence; birthday problem

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# 1. Introduction

The classic coupon collector problem has a long history; see, for example, [3]. The classic problem is as follows. A collector wishes to collect a complete set of n distinct coupons, labelled 1 through to n. The coupons are hidden inside breakfast cereal boxes and within each cereal box there is one coupon which is equally likely to be any of the n distinct coupons. The collector purchases one box of breakfast cereal at a time, collecting the coupons, stopping when the collector has completed the set of n distinct coupons. The total number of cereal boxes,  $Y_n$ , which the collector needs to purchase is the quantity of interest. Elementary calculations show that

$$E[Y_n] = n \sum_{i=1}^n \frac{1}{i} \approx n \log n.$$

Furthermore, if Z is a standard Gumbel distribution with  $P(Z \le z) = \exp(-e^{-z})$   $(z \in \mathbb{R})$  then

$$\frac{1}{n}(Y_n - n\log n) \stackrel{\mathrm{D}}{\to} Z \quad \text{as } n \to \infty,$$

where  $\stackrel{\text{D}}{\rightarrow}$  denotes convergence in distribution; see, for example, [4].

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The generalised coupon collector problem assumes that, whilst the cereal boxes are independent and identically distributed, the probability that a box contains coupon i is  $p_i$ . No assumption is placed upon the  $\{p_i\}$ s except that  $p_i > 0$  (i = 1, 2, ..., n). We allow for the possibility that some boxes may not contain a coupon by only assuming that  $\sum_{i=1}^{n} p_i \le 1$ . The random coupon collector problem [4], [5] is an alternative departure from the classic problem. The proofs in [4] rely upon a Poisson embedding argument and although our proofs are different we shall also exploit a Poisson approximation approach.

The paper is structured as follows. In Section 2 the main result, Theorem 2.1, is presented and proved. An alternative result is given in Theorem 2.2 which is applicable when the Poisson arguments of Theorem 2.1 fail. A number of examples are considered in Section 3. Finally, in Section 4 extensions of Section 2 are discussed. These include the K-coupon collector problem, the total number of draws from the population that are required to have K coupons of each type, and the K-birthday problem, the total number of draws from the population that are required to have K coupons of any (unspecified) type.

## 2. Coupon collecting problem

For the asymptotic results of this paper, we consider a sequence of coupon collections  $\{C_n\}$ , where the number of coupons to be collected, n, tends to  $\infty$ . For  $n \ge 1$ ,  $C_n$  requires the collection of n coupons, labelled 1 through to n. Coupons are collected as follows. Let  $X_1^n, X_2^n, \ldots$  be independent and identically distributed according to  $X^n$ , where

$$P(X^n = i) = \begin{cases} p_{ni}, & i = 1, 2, \dots, n, \\ 0, & \text{otherwise,} \end{cases}$$

where  $\sum_{i=1}^{n} p_{ni} \le 1$  and  $\min_{1 \le i \le n} p_{ni} > 0$ . Then  $X_k^n$  is the kth coupon drawn from the population (of coupons) and the process is continued until all n coupons have been collected. Let  $Y_n$  denote the total number of coupons which need to be collected to obtain the full set of coupons in  $\mathcal{C}_n$ .

Before stating the main result, we introduce some useful notation. For  $n \ge 1$ , i = 1, 2, ..., n, and t = 1, 2, ..., let  $\chi_i^n(t) = 1$  if coupon i has not been collected in the first t coupons drawn from the population and  $\chi_i^n(t) = 0$  otherwise. Let  $W_n(t) = \sum_{i=1}^n \chi_i^n(t)$ , the total number of distinct coupons which still need to be collected after t coupon draws. Thus, for  $t \ge 1$ ,  $Y_n \le t$  if and only if  $W_n(t) = 0$ .

**Theorem 2.1.** Suppose that there exist sequences  $\{b_n\}$  and  $\{k_n\}$  such that  $k_n/b_n \to 0$  as  $n \to \infty$  and that, for  $y \in \mathbb{R}$ ,

$$\sum_{i=1}^{n} \exp(-p_{ni}\{b_n + yk_n\}) \to g(y) \quad as \ n \to \infty$$
 (2.1)

for a nonincreasing function  $g(\cdot)$  with  $g(y) \to \infty$  as  $y \to -\infty$  and  $g(y) \to 0$  as  $y \to \infty$ . Then, if  $\tilde{Y}_n = (Y_n - b_n)/k_n$ ,

$$\tilde{Y}_n \stackrel{\mathrm{D}}{\to} Y \quad as \ n \to \infty,$$

where Y has cumulative distribution function

$$P(Y \le y) = e^{-g(y)}, \quad y \in \mathbb{R}.$$

The key restriction in Theorem 2.1 is that (2.1) implies that  $\min_{1 \le i \le n} p_{ni}b_n \to \infty$  as  $n \to \infty$ . This condition is needed for the Poisson limit (2.2), below, since it implies that  $\max_{1 \le i \le n} \mathbb{E}[\chi_i^n([b_n + yk_n])] \to 0$  as  $n \to \infty$ . In Theorem 2.2, below, we explore the case where  $\min_{1 \le i \le n} p_{ni}b_n \to c$  as  $n \to \infty$  for some  $0 < c < \infty$ . By Jensen's inequality,

$$\sum_{i=1}^{n} \exp(-p_{ni}b_n) \ge \sum_{i=1}^{n} \exp\left(-\frac{1}{n}b_n\right) = n \exp\left(-\frac{b_n}{n}\right).$$

Therefore,  $b_n \ge n \log n$ , and this will be used in Lemma 2.2, below. The only restriction placed upon the sequence  $\{X^n\}$  is (2.1). Discussion of a natural construction of suitable sequences  $\{X^n\}$  is deferred to Section 3.

The proof of Theorem 2.1 relies upon two preliminary lemmas which are motivated and proved in the following discussion.

Since, for  $t \ge 1$ ,  $Y_n \le t$  if and only  $W_n(t) = 0$ , it suffices to show that, for all  $y \in \mathbb{R}$ ,

$$W_n([b_n + yk_n]) \xrightarrow{D} Po(g(y)), \quad y \in \mathbb{R}.$$
 (2.2)

The first step in proving (2.2) is to show that, for any  $t \in \mathbb{N}$ ,  $\{\chi_i^n(t)\}$  are negatively related [1, p. 24]. For  $n, t \ge 1$  and  $1 \le j \le n$ , let  $\{\theta_{i,j}^n(t); i = 1, 2, ..., n\}$  be random variables satisfying

$$\mathcal{L}(\theta_{i,j}^n(t); i = 1, 2, \dots, n) = \mathcal{L}(\chi_i^n(t); i = 1, 2, \dots, n \mid \chi_j^n(t) = 1).$$

**Lemma 2.1.** For  $n, t \ge 1$ , the random variables  $\{\chi_i^n(t)\}$  are negatively related, i.e. for each  $1 \le j \le n$ , the random variables  $\{\theta_{i,j}^n(t); i = 1, 2, ..., n\}$  and  $\{\chi_i^n(t); i = 1, 2, ..., n\}$  can be defined on a common probability space  $(\Omega, \mathcal{F}, P)$  such that, for all  $i \ne j$ ,  $\chi_i^n(t)(\omega) \ge \theta_{i,j}^n(t)(\omega)$  for all  $\omega \in \Omega$ .

*Proof.* The lemma is proved by a simple coupling argument.

Fix  $n, t \ge 1$  and j = 1, 2, ..., n. Draw  $X_1^n, X_2^n, ..., X_t^n$  from  $X^n$ . For k = 1, 2, ..., t, let  $\tilde{X}_k^n(t) \stackrel{\text{D}}{=} X_k^n \mid \chi_j^n(t) = 1$ , where ' $\stackrel{\text{D}}{=}$ ' denotes equality in distribution. For k = 1, 2, ..., t, if  $X_k^n \ne j$ , set  $\tilde{X}_k^n(t) = X_k^n$ . If  $X_k^n = j$ , set  $\tilde{X}_k^n(t) = \hat{X}_k^n$ , where

$$P(\hat{X}_k^n = i) = \begin{cases} \frac{p_{ni}}{1 - p_{nj}}, & i \neq j, \\ 0, & \text{otherwise.} \end{cases}$$

Thus,  $\tilde{X}_1^n(t)$ ,  $\tilde{X}_2^n(t)$ , ...,  $\tilde{X}_t^n(t)$  have the correct distribution and, by construction,  $\chi_i^n(t) \ge \theta_{i,j}^n(t)$  for  $i \ne j$ .

Note that

$$E[W_n([b_n + yk_n])] = \sum_{i=1}^n (1 - p_{ni})^{[b_n + yk_n]} \to g(y) \quad \text{as } n \to \infty.$$

Therefore, by Lemma 2.1 and [1, Corollary 2.C.2], (2.2) holds if

$$\operatorname{var}(W_n([b_n + yk_n])) \to g(y) \quad \text{as } n \to \infty.$$
 (2.3)

Now  $var(W_n([b_n + yk_n]))$  is equal to

$$\sum_{i=1}^{n} \operatorname{var}(\chi_{i}^{n}([b_{n} + yk_{n}])) + \sum_{i=1}^{n} \sum_{j \neq i} \operatorname{cov}(\chi_{i}^{n}([b_{n} + yk_{n}]), \chi_{j}^{n}([b_{n} + yk_{n}])). \tag{2.4}$$

Equation (2.1) ensures that

$$\sum_{i=1}^{n} \exp(-p_{ni}[b_n + yk_n])^2 \to 0 \quad \text{as } n \to \infty.$$

Therefore, by (2.1), the first term in (2.4) converges to g(y) as  $n \to \infty$ . Thus, (2.3) holds if the latter term in (2.4) converges to 0 as  $n \to \infty$ .

## Lemma 2.2.

$$\sum_{i=1}^n \sum_{j\neq i} |\operatorname{cov}(\chi_i^n([b_n + yk_n]), \chi_j^n([b_n + yk_n]))| \to 0 \quad \text{as } n \to \infty.$$

*Proof.* For any  $i \neq j$ ,

$$\begin{aligned} &|\operatorname{cov}(\chi_{i}^{n}([b_{n}+yk_{n}]),\chi_{j}^{n}([b_{n}+yk_{n}]))| \\ &= |(1-p_{ni}-p_{nj})^{[b_{n}+yk_{n}]} - (1-p_{ni})^{[b_{n}+yk_{n}]}(1-p_{nj})^{[b_{n}+yk_{n}]}| \\ &= (1-p_{ni})^{[b_{n}+yk_{n}]}(1-p_{nj})^{[b_{n}+yk_{n}]} \left| \left(1 - \frac{p_{ni}p_{nj}}{(1-p_{ni})(1-p_{nj})}\right)^{[b_{n}+yk_{n}]} - 1 \right| \\ &\leq (1-p_{ni})^{[b_{n}\log n + yn]}(1-p_{nj})^{[b_{n}+yk_{n}]} \frac{[b_{n}+yk_{n}]p_{ni}p_{nj}}{(1-p_{ni})(1-p_{nj})}, \end{aligned}$$

with the inequality coming from  $|1 - (1 - y)^m| \le my$  for  $0 \le y \le 1$  and  $m \in \mathbb{N}$ . Therefore,

$$\sum_{i=1}^{n} \sum_{j \neq i} |\text{cov}(\chi_{i}^{n}([b_{n} + yk_{n}]), \chi_{j}^{n}([b_{n} + yk_{n}]))|$$

$$\leq \left(\sqrt{[b_{n} + yk_{n}]} \sum_{i=1}^{n} \frac{p_{ni}}{1 - p_{ni}} (1 - p_{ni})^{[b_{n} + yk_{n}]}\right)^{2}.$$
(2.5)

Let  $A_n = \{i; p_{ni} \le b_n^{-3/4}\}$ . Then

$$\sqrt{[b_n + yk_n]} \sum_{i=1}^n \frac{p_{ni}}{1 - p_{ni}} (1 - p_{ni})^{[b_n + yk_n]} 
= \sqrt{[b_n + yk_n]} \sum_{i \in \mathcal{A}_n} \frac{p_{ni}}{1 - p_{ni}} (1 - p_{ni})^{[b_n + yk_n]} 
+ \sqrt{[b_n + yk_n]} \sum_{i \in \mathcal{A}_n^c} \frac{p_{ni}}{1 - p_{ni}} (1 - p_{ni})^{[b_n + yk_n]} 
\leq \frac{b_n^{-3/4} \sqrt{[b_n + yk_n]}}{1 - b_n^{-3/4}} \sum_{i=1}^n (1 - p_{ni})^{[b_n + yk_n]} + \sqrt{[b_n + yk_n]} \sum_{i \in \mathcal{A}_n^c} (1 - b_n^{-3/4})^{[b_n + yk_n] - 1} 
\to 0 \quad \text{as } n \to \infty,$$

since  $\sum_{i=1}^{n} (1 - p_{ni})^{[b_n + yk_n]} \to g(y)$  and  $b_n \ge n \log n$  as  $n \to \infty$ . Therefore, the right-hand side of (2.5) converges to 0 as  $n \to \infty$  and the lemma is proved.

*Proof of Theorem 2.1.* For any  $y \in \mathbb{R}$ ,  $\tilde{Y}_n \leq y$  if and only if  $W_n([b_n + yk_n]) = 0$ . Therefore, by (2.2), for  $y \in \mathbb{R}$ ,

$$P(\tilde{Y}_n \le y) = P(W_n([b_n + yk_n]) = 0)$$

$$\to e^{-g(y)}$$

$$= P(Y \le y) \text{ as } n \to \infty,$$

and the theorem is proved.

The proof of Theorem 2.1 presents a straightforward bound for  $|P(\tilde{Y}_n \leq y) - P(Y \leq y)|$ ,  $y \in \mathbb{R}$ . For  $t \geq 0$ , let  $Z(t) \sim Po(t)$  and, for  $y \in \mathbb{R}$ , let  $g_n(y) = E[W_n([b_n + yk_n])]$ . By the triangle inequality and [1, Corollary 2.C.2],

$$\begin{split} |\mathrm{P}(\tilde{Y}_n \leq y) - \mathrm{P}(Y \leq y)| \\ &= |\mathrm{P}(W_n([b_n + yk_n]) = 0) - \mathrm{P}(Z(g(y)) = 0)| \\ &\leq |\mathrm{P}(W_n([b_n + yk_n]) = 0) - \mathrm{P}(Z(g_n(y)) = 0)| + |\mathrm{P}(Z(g_n(y)) = 0) - \mathrm{P}(Z(g(y)) = 0)| \\ &\leq (1 - \exp(-g_n(y))) \left(1 - \frac{\mathrm{var}(W_n([b_n + yk_n]))}{g_n(y)}\right) + |\exp(-g_n(y)) - \mathrm{e}^{-g(y)}|. \end{split}$$

We now turn our attention to the situation where the natural scaling  $\{b_n\}$  is such that  $\min_{1 \le i \le n} p_{ni}b_n \to c$  as  $n \to \infty$  for some  $0 < c < \infty$ .

**Theorem 2.2.** Suppose that there exist sequences  $\{b_n\}$  such that, for  $y \in \mathbb{R}^+$ ,

$$\sum_{i=1}^{n} \exp(-p_{ni}yb_n) \to g(y) \quad as \ n \to \infty$$
 (2.6)

for a nonincreasing function  $g(\cdot)$  with  $g(y) \to \infty$  as  $y \to 0$  and  $g(y) \to 0$  as  $y \to \infty$ . Suppose that there exists a function  $h(\cdot)$  such that, for all  $y \in \mathbb{R}^+$ ,

$$\prod_{i=1}^{n} (1 - \exp(-p_{ni}yb_n)) \to h(y) \quad as \ n \to \infty.$$

Then (2.6) ensures that  $h(y) \to 0$  as  $y \to 0$  and  $h(y) \to 1$  as  $y \to \infty$ , and if  $\hat{Y}_n = Y_n/b_n$ ,

$$\hat{Y}_n \stackrel{\mathrm{D}}{\to} Y \quad as \ n \to \infty,$$

where Y has cumulative distribution function

$$P(Y < y) = h(y), \quad y \in \mathbb{R}^+.$$

*Proof.* The proof has a number of similarities and differences to the proof of Theorem 2.1. We shall again exploit the fact that  $Y_n \le t$  if and only  $W_n(t) = 0$ .

Let  $\eta_*^n$  be a homogeneous Poisson point process with rate 1, and let  $T_n(t)$  denote the time of the  $[tb_n]$ th point on  $\eta_*^n$ . Let  $V_1^n, V_2^n, \ldots$  be independent and identically distributed according to  $X^n$ . Let  $\eta_1^n, \eta_2^n, \ldots, \eta_n^n$  be independent homogeneous Poisson point processes with rates

 $p_{n1}, p_{n2}, \ldots, p_{nn}$ , respectively, constructed from  $\eta_*^n$  and  $V_1^n, V_2^n, \ldots$  as follows. For  $k = 1, 2, \ldots$ , let  $s_k^n$  denote the time of the kth point on  $\eta_*^n$ . Then there is a point on  $\eta_j^n$  at time  $s_k^n$  if  $V_k^n = j$ . Furthermore,  $\chi_1^n(t), \chi_2^n(t), \ldots, \chi_n^n(t)$  and, hence,  $W_n(t)$  can be constructed using  $V_1^n, V_2^n, \ldots, V_t^n$ .

Let  $\psi_i^n(t) = 1$  if there is no point on  $\eta_i^n[0, t]$ , and note that the  $\{\psi_i^n(t)\}$ s are independent. For  $t \ge 0$ , let  $\tilde{W}_n(t) = \sum_{i=1}^n \psi_i^n(t)$ . Then  $W_n([yb_n]) = \tilde{W}_n(T_n([yb_n]))$ . Since  $\tilde{W}_n(\cdot)$  is nondecreasing, if  $[yb_n] - ([yb_n])^{3/4} \le T_n([yb_n]) \le [yb_n] + ([yb_n])^{3/4}$  then

$$\tilde{W}_n([yb_n] + ([yb_n])^{3/4}) \le W_n([yb_n]) \le \tilde{W}_n([yb_n] - ([yb_n])^{3/4}). \tag{2.7}$$

Since  $(1/(yb_n)^{3/4})(T_n([yb_n]) - [yb_n]) \stackrel{P}{\to} 0$  as  $n \to \infty$  (where ' $\stackrel{P}{\to}$ ' denotes convergence in probability), it follows from (2.7) that  $P(W_n([yb_n]) = 0) \to h(y)$  if

$$P(\tilde{W}_n([yb_n] \pm ([yb_n])^{3/4}) = 0) \to h(y)$$
 as  $n \to \infty$ .

By independence, for all  $y \in \mathbb{R}$ ,

$$P(\tilde{W}_n([yb_n] \pm ([yb_n])^{3/4}) = 0) = \prod_{i=1}^n (1 - (1 - p_{ni})^{[yb_n] \pm ([yb_n])^{3/4}})$$

$$\to h(y) \quad \text{as } n \to \infty.$$

The main benefit of Theorem 2.1 over Theorem 2.2 is that g(y) is usually much easier to calculate than h(y).

### 3. Examples

A natural construction of  $\{X^n\}$  is to take a (continuous) distribution X with probability density function  $f(\cdot)$  on [0, 1] and, for  $n = 1, 2, \ldots$  and  $i = 1, 2, \ldots, n$ , set

$$p_{ni} = \int_{(i-1)/n}^{i/n} f(x) \, \mathrm{d}x.$$

A number of results can be proved concerning various choices of X with Lemma 3.1 illustrating the point using a class of distributions with  $f(\cdot)$  being continuous.

**Lemma 3.1.** Let  $0 \le \sigma \le 1$  be such that, for all  $0 \le x \le 1$  and  $x \ne \sigma$ ,  $0 < f(\sigma) < f(x)$ . For p = 1, 2, let

$$u_p = \lim_{\varepsilon \to 0+} \frac{f(\sigma + \varepsilon) - f(\sigma)}{\varepsilon^p}, \qquad l_p = \lim_{\varepsilon \to 0-} \frac{f(\sigma + \varepsilon) - f(\sigma)}{|\varepsilon|^p}.$$

(i) Suppose that  $\mathbf{1}_{\{\sigma>0\}} l_1 + \mathbf{1}_{\{\sigma<1\}} u_1 > 0$ . Then  $b_n = (n/f(\sigma))(\log n - \log(\log n))$  and  $k_n = n$  with

$$g(y) = f(\sigma) \left( \frac{\mathbf{1}_{\{\sigma > 0\}}}{l_1} + \frac{\mathbf{1}_{\{\sigma < 1\}}}{u_1} \right) e^{-f(\sigma)y}.$$

(ii) Suppose that  $\mathbf{1}_{\{\sigma>0\}} l_1 + \mathbf{1}_{\{\sigma<1\}} u_1 = 0$  and  $\mathbf{1}_{\{\sigma>0\}} l_2 + \mathbf{1}_{\{\sigma<1\}} u_2 > 0$ . Then  $b_n = (n/f(\sigma))(\log n - \frac{1}{2}\log(\log n))$  and  $k_n = n$  with

$$g(y) = \sqrt{\frac{\pi f(\sigma)}{2}} \left( \sqrt{\frac{\mathbf{1}_{\{\sigma > 0\}}}{l_2}} + \sqrt{\frac{\mathbf{1}_{\{\sigma < 1\}}}{u_2}} \right) e^{-f(\sigma)y}.$$

*Proof.* We outline the proof of (i), with (ii) being proved similarly. Let  $b_n = (n/f(\sigma))(\log n - \log(\log n))$  and  $k_n = n$ . Note that

$$\sum_{i=1}^{n} \exp(-p_{ni}(b_n + yk_n)) \approx \sum_{i=1}^{n} \exp\left(-(b_n + yk_n)\frac{1}{n}f\left(\frac{i - 1/2}{n}\right)\right)$$
$$= n\sum_{i=1}^{n} \frac{1}{n} \exp\left(-\left(\frac{b_n}{n} + y\right)f\left(\frac{i - 1/2}{n}\right)\right)$$
$$\approx n\int_0^1 \exp\left(-\left(\frac{b_n}{n} + y\right)f(x)\right) dx.$$

Therefore, it is straightforward to show that

$$g(y) = \lim_{n \to \infty} n \int_0^1 \exp\left(-\left(\frac{1}{f(\sigma)}(\log n - \log(\log n)) + y\right)f(x)\right) dx.$$

Linearizing f(x) about  $\sigma$  and considering the left- and right-hand limits separately yields the result.

Examples of probability density functions on [0, 1] satisfying Lemma 3.1 include  $f(x) = \frac{2}{3}(1+x)$ ,  $f(x) = \frac{6}{5}(1-x(1-x))$ , and  $f(x) = \frac{12}{7}\max(1-x,x/2)$ .

Suppose instead that X is piecewise constant with, for  $1 \le j \le k$ ,

$$f(x) = \lambda_j, \qquad \pi_{j-1} < x \le \pi_j,$$

where  $\lambda_1, \lambda_2, \dots, \lambda_k > 0$  and  $0 = \pi_0 < \pi_1 < \dots < \pi_k = 1$ . Without loss of generality, assume that  $\lambda_1 < \lambda_2 < \dots < \lambda_k$ . Then  $b_n = (1/\lambda_1)n \log n$ ,  $k_n = n$ , and  $g(y) = \pi_1 \exp(-\lambda_1 y)$ .

In the above examples,  $k_n/b_n \to 0$  and Theorem 2.1 holds. In all cases, the limiting distribution Y is a Gumbel distribution with  $b_n/n \log n \to 1/\min_{0 \le x \le 1} f(x)$  as  $n \to \infty$ .

An example of where Theorem 2.2 is necessary is f(x) = 2x  $(0 \le x \le 1)$ , giving  $p_{ni} = (2i-1)/n^2$  (i = 1, 2, ..., n). Then, for  $y \in \mathbb{R}^+$ ,

$$\sum_{i=1}^{n} \exp(-p_{ni} y n^2) = \sum_{i=1}^{n} \exp(-(2i-1)y) \to g(y) = \frac{e^y}{e^{2y} - 1} \quad \text{as } n \to \infty,$$

and Theorem 2.2 holds with  $b_n = n^2$  and  $h(y) = \lim_{n \to \infty} \prod_{i=1}^n (1 - \exp(-(2i - 1)y))$ .

#### 4. Extensions

The methodology outlined in Section 2 can be extended to find the total number of coupons,  $Y_n^K$ , which need to be collected in order to have (at least) K coupons of each type. In this case, simply let  $\chi_i^n(t) = 1$  if at most K - 1 coupons of type i have been collected in the first t draws from the population and let  $\chi_i^n(t) = 0$  otherwise. Then set  $W_n^K(t) = \sum_{i=1}^n \chi_i^n(t)$ , and note that  $Y_n^K \le t$  if and only if  $W_n^K(t) = 0$ . It is straightforward to adapt Lemmas 2.1 and 2.2 to this case and, consequently, Theorem 2.1 holds with (2.1) replaced by

$$\frac{b_n^{K-1}}{(K-1)!} \sum_{i=1}^n p_{ni}^{K-1} \exp(-p_{ni} \{b_n + yk_n\}) \to g(y) \quad \text{as } n \to \infty.$$
 (4.1)

Since  $k_n/b_n \to 0$  implies that  $\min_{1 \le i \le n} b_n p_{ni} \to \infty$  as  $n \to \infty$ , (4.1) holds if and only if

$$E[W_n^K([b_n + yk_n])] \to g(y)$$
 as  $n \to \infty$ .

Theorem 2.2 can also be adapted to the *K*-coupon collector problem.

At the other end of the spectrum, the Poisson arguments above can be applied to the generalised birthday problem. That is, for  $K \geq 2$ , let  $U_n^K$  denote the total number of draws from the population that are required to obtain K coupons of any (unspecified) type. Let  $\tilde{\chi}_i^n(t) = 1$  if at least K coupons of type i have been collected in the first t draws from the population and let  $\tilde{\chi}_i^n(t) = 0$  otherwise. Then, if  $\tilde{W}_n^K(t) = \sum_{i=1}^n \tilde{\chi}_i^n(t)$ ,  $U_n^K > t$  if and only if  $\tilde{W}_n^K(t) = 0$ . Along the lines of Lemma 2.1, it can be shown that the  $\{\tilde{\chi}_i^n(t)\}$  are negatively related and straightforward bounds for the covariance terms can be obtained. We then have the following theorem.

**Theorem 4.1.** For fixed  $K \geq 2$ , suppose that there exists a sequence  $\{l_n\}$  such that

$$l_n^K \sum_{i=1}^n p_{ni}^K \to 1$$
 (4.2)

and  $\max_{1 \le i \le n} l_n p_{ni} \to 0$  as  $n \to \infty$ . Then

$$\frac{U_n^K}{l_n} \xrightarrow{\mathrm{D}} U^K \quad as \ n \to \infty,$$

where  $U^K$  has cumulative distribution function

$$P(U^K \le u) = 1 - \exp(-u^K), \quad u \in \mathbb{R}^+.$$

*Proof.* The conditions imposed on  $\{l_n\}$  are sufficient for  $W_n^K([ul_n]) \xrightarrow{D} Po(u^K)$ , from which the theorem follows immediately.

The limiting distribution  $U^K$  obtained in Theorem 4.1 is identical to that obtained in [4, Theorem 5.2] for the random birthday problem. For the case in which K=2, Theorem 4.1 follows immediately from [2, Example 2], since given (4.2),  $\max_{1 \le i \le n} l_n p_{ni} \to 0$  if and only if  $l_n^3 \sum_{i=1}^n p_{ni}^3 \to 0$  as  $n \to \infty$ .

Finally, it is worth noting that, for the establishing of Poisson limits for  $W_n^K([b_n + yk_n])$  and  $\tilde{W}_n^K([ul_n])$ , it is crucial that

$$\max_{1 \le i \le n} \mathbb{E}[\chi_i^n([b_n + yk_n])] \to 0 \quad \text{and} \quad \max_{1 \le i \le n} \mathbb{E}[\tilde{\chi}_i^n([ul_n])] \to 0 \quad \text{as } n \to \infty,$$

respectively. That is, for the K-coupon collector problem, we require that  $\min_{1 \le i \le n} b_n p_{ni} \to \infty$  as  $n \to \infty$  (none of the probabilities are too small) and, for the K-birthday problem, we require that  $\max_{1 \le i \le n} l_n p_{ni} \to 0$  as  $n \to \infty$  (none of the probabilities are too large).

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