

# Min-Wise independent linear permutations

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January 13, 2000

## 1 Introduction

Broder, Charikar, Frieze and Mitzenmacher [3] introduced the notion of a set of min-wise independent permutations. We say that  $\mathcal{F} \subseteq S_n$  is *min-wise independent* if for any set  $X \subseteq [n]$  and any  $x \in X$ , when  $\pi$  is chosen at random in  $\mathcal{F}$  we have

$$\mathbb{P}(\min\{\pi(X)\} = \pi(x)) = \frac{1}{|X|}. \quad (1)$$

The research was motivated by the fact that such a family (under some relaxations) is essential to the algorithm used in practice by the AltaVista web index software to detect and filter near-duplicate documents. A set of permutations satisfying (1) needs to be exponentially large [3]. In practice we can allow certain relaxations. First, we can accept small relative errors. We say that  $\mathcal{F} \subseteq S_n$  is *approximately min-wise independent with relative error  $\epsilon$*  (or just approximately min-wise independent, where the meaning is clear) if for any set  $X \subseteq [n]$  and any  $x \in X$ , when  $\pi$  is chosen at random in  $\mathcal{F}$  we have

$$\left| \mathbb{P}(\min\{\pi(X)\} = \pi(x)) - \frac{1}{|X|} \right| \leq \frac{\epsilon}{|X|}. \quad (2)$$

In other words we require that all the elements of any fixed set  $X$  have only an almost equal chance to become the minimum element of the image of  $X$  under  $\pi$ .

*Linear permutations* are an important class of permutations. Let  $p$  be a (large) prime and let  $\mathcal{F}_p = \{\pi_{a,b} : 1 \leq a \leq p-1, 0 \leq b \leq p-1\}$  where for  $x \in [p] = \{0, 1, \dots, p-1\}$ ,

$$\pi_{a,b}(x) = ax + b \pmod{p},$$

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where for integer  $n$  we define  $n \bmod p$  to be the non-negative remainder on division of  $n$  by  $p$ .

For  $X \subseteq [p]$  we let

$$F(X) = \max_{x \in X} \{\mathbb{P}_{a,b}(\min\{\pi(X)\} = \pi(x))\}$$

where  $\mathbb{P}_{a,b}$  is over  $\pi$  chosen uniformly at random from  $\mathcal{F}_p$ . The natural questions to discuss are what are the extremal and average values of  $F(X)$  as  $X$  ranges over  $\mathcal{A}_k = \{X \subseteq [p] : |X| = k\}$ . The following results were some of those obtained in [3]:

**Theorem 1**

(a) Consider the set  $X_k = \{0, 1, 2 \dots k-1\}$ , as a subset of  $[p]$ . As  $k, p \rightarrow \infty$ , with  $k^2 = o(p)$ ,

$$\mathbb{P}_{a,b}(\min\{\pi(X_k)\} = \pi(0)) = \frac{3 \ln k}{\pi^2 k} + O\left(\frac{k^2}{p} + \frac{1}{k}\right).$$

(b) As  $k, p \rightarrow \infty$ , with  $k^4 = o(p)$ ,

$$\frac{1}{2(k-1)} \leq \mathbb{E}_X[F(X)] \leq \frac{\sqrt{2}+1}{\sqrt{2}k} + O\left(\frac{1}{k^2}\right),$$

where  $\mathbb{E}_X$  denotes expectations over  $X$  chosen uniformly at random from  $\mathcal{A}_k$ .

In this paper we improve the second result and prove

**Theorem 2**

As  $k, p \rightarrow \infty$ ,

$$\mathbb{E}_X[F(X)] = \frac{1}{k} + O\left(\frac{(\log k)^3}{k^{3/2}}\right).$$

Thus for most sets, simply chosen, random linear permutations, will suffice as (near) min-wise independent. Other results on min-wise independence have been obtained by Indyk [6], Broder, Charikar and Mitzenmacher [4] and Broder and Feige [5].

## 2 Proof of Theorem 2

Let  $X = \{x_0, x_1, \dots, x_{k-1}\} \subseteq [p]$ . Let  $\beta_i = ax_i \bmod p$  for  $i = 0, 1, \dots, k-1$ . Let

$$i = i(X, a) = \min\{\beta_0 - \beta_j \bmod p : j = 1, 2, \dots, k-1\}. \tag{3}$$

Let

$$A_i = A_i(X) = \{a \in [p] : i(X, a) = i\}$$

and note that

$$|A_i| \leq k - 1, \quad i = 1, 2, \dots, p - 1.$$

Then

$$\min\{\pi(X)\} = \pi(x_0) \text{ iff } 0 \in \{\beta_0 + b, \beta_0 + b - 1, \dots, \beta_0 + b - i + 1\} \bmod p.$$

Thus if

$$Z = Z(X) = \sum_{i=1}^{p-1} i|A_i|,$$

$$\mathbb{P}_{a,b}(\min\{\pi(X)\} = \pi(x_0)) = \frac{Z}{p(p-1)}. \quad (4)$$

Fix  $a \in \{1, 2, \dots, p-1\}$  and  $x_0$ . Then

$$\mathbb{P}(a \in A_i) = (k-1) \cdot \frac{1}{p-1} \prod_{t=1}^{k-2} \left(1 - \frac{i+t}{p-1-t}\right) \quad (5)$$

We write  $Z = Z_0 + Z_1$  where  $Z_0 = \sum_{i=1}^{i_0} i|A_i|$  where  $i_0 = \frac{4p \log k}{k}$ . Now, by symmetry,

$$\mathbb{E}_X(\mathbb{P}_{a,b}(\min\{\pi(X)\} = \pi(x_0))) = \frac{1}{k} \quad (6)$$

and so

$$\mathbb{E}_X(Z) = \frac{p(p-1)}{k}.$$

It follows from (5) that

$$\begin{aligned} \mathbb{E}(Z_1) &\leq (k-1) \sum_{i=i_0+1}^{p-1} i \exp\left\{-\frac{4(k-2) \log k}{k}\right\} \\ &\leq \frac{p^2}{k^3} \end{aligned} \quad (7)$$

for large  $k, p$ .

We continue by using the Azuma-Hoeffding Martingale tail inequality – see for example [1, 2, 7, 8, 9]. Let  $x_0$  be fixed and for a given  $X$  let  $\hat{X}$  be obtained from  $X$  by replacing  $x_j$  by randomly chosen  $\hat{x}_j$ . For  $j \geq 1$  let

$$d_j = \max_X \{|\mathbb{E}_{\hat{x}_j}(Z(X) - Z(\hat{X}))|\}.$$

Then for any  $t > 0$  we have

$$\mathbb{P}(|Z_0 - \mathbb{E}(Z_0)| \geq t) \leq 2 \exp\left\{-\frac{2t^2}{d_1^2 + \dots + d_{k-1}^2}\right\}. \quad (8)$$

We claim that

$$d_j \leq \sum_{i=1}^{i_0} i + \sum_{i=1}^{i_0} \frac{(k-1)i^2}{p} \quad (9)$$

$$\begin{aligned} &\leq \frac{i_0^2}{2} + \frac{i_0^3 k}{3p} + O(p) \\ &\leq \frac{30(\log k)^3 p^2}{k^2} \end{aligned} \quad (10)$$

**Explanation for (9):** If  $a \in A_i(X)$  because  $ax_j = ax_0 - i \pmod p$  then changing  $x_j$  to  $\hat{x}_j$  changes  $|A_i|$  by one. This explains the first summation. The second accounts for those  $a \in A_i(X)$  for which  $ax_0 - a\hat{x}_j \pmod p < i$ , changing the minimum in (3). We then use  $|A_i| \leq k-1$  and  $\mathbb{P}(ax_0 - a\hat{x}_j \pmod p < i) = \frac{i}{p}$ .

Using (10) in (8) with  $t = \varepsilon \frac{p^2}{k}$  we see that

$$\mathbb{P}\left(|Z_0 - \mathbb{E}(Z_0)| \geq \varepsilon \frac{p^2}{k}\right) \leq \exp\left\{-\frac{\varepsilon^2 k}{450(\log k)^6}\right\}.$$

It now follows from (4), (6), (7) and the above that

$$\mathbb{E}_X[F(X)] = \frac{1}{k} + O\left(\frac{1}{k^2} + \frac{1}{k} \int_{\varepsilon=0}^{\infty} \min\left\{1, k \exp\left\{-\frac{\varepsilon^2 k}{450(\log k)^6}\right\}\right\} d\varepsilon\right)$$

and the result follows. □

## References

- [1] N.Alon and J.H.Spencer, *The Probabilistic Method*, Wiley, 1992.
- [2] B.Bollobás, *Martingales, isoperimetric inequalities and random graphs*, in Combinatorics, A.Hajnal, L.Lovász and V.T.Sós Ed., Colloq. Math. Sci. Janos Bolyai 52, North Holland 1988.
- [3] A.Z.Broder, M.Charikar, A.M.Frieze and M.Mitzenmacher, *Min-Wise Independent permutations*
- [4] A.Z.Broder, M.Charikar and M.Mitzenmacher, *A derandomization using min-wise independent permutations*, Proceedings of Second International Workshop RANDOM '98 (M.Luby, J.Rolim, M.Serna Eds.) (1998) 15-24.
- [5] A.Z.Broder and U.Feige, *Min-Wise versus Linear Independence*, Proceedings of the 11th Annual ACM-SIAM Symposium on Discrete Algorithms (2000).

- [6] P.Indyk, *A small approximately min-wise independent family of hash-functions*, Proceedings of the 10th Annual ACM-SIAM Symposium on Discrete Algorithms (1999) 454-456.
- [7] C.J.H.McDiarmid, *On the method of bounded differences*, Surveys in Combinatorics, 1989, Invited papers at the Twelfth British Combinatorial Conference, Edited by J.Siemons, Cambridge University Press, 148-188.
- [8] C.J.H.McDiarmid, *Concentration*, Probabilistic methods for algorithmic discrete mathematics, (M.Habib, C.McDiarmid, J.Ramirez-Alfonsin, B.Reed, Eds.), Springer (1998) 195-248.
- [9] M.J.Steele, *Probability theory and combinatorial optimization*, CBMS-NSF Regional Conference Series in Applied Mathematics 69, 1997.