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## Selected Solutions for Chapter 5: Probabilistic Analysis and Randomized Algorithms

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### Solution to Exercise 5.2-1

Since HIRE-ASSISTANT always hires candidate 1, it hires exactly once if and only if no candidates other than candidate 1 are hired. This event occurs when candidate 1 is the best candidate of the  $n$ , which occurs with probability  $1/n$ .

HIRE-ASSISTANT hires  $n$  times if each candidate is better than all those who were interviewed (and hired) before. This event occurs precisely when the list of ranks given to the algorithm is  $\langle 1, 2, \dots, n \rangle$ , which occurs with probability  $1/n!$ .

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### Solution to Exercise 5.2-4

Another way to think of the hat-check problem is that we want to determine the expected number of fixed points in a random permutation. (A *fixed point* of a permutation  $\pi$  is a value  $i$  for which  $\pi(i) = i$ .) We could enumerate all  $n!$  permutations, count the total number of fixed points, and divide by  $n!$  to determine the average number of fixed points per permutation. This would be a painstaking process, and the answer would turn out to be 1. We can use indicator random variables, however, to arrive at the same answer much more easily.

Define a random variable  $X$  that equals the number of customers that get back their own hat, so that we want to compute  $E[X]$ .

For  $i = 1, 2, \dots, n$ , define the indicator random variable

$$X_i = I\{\text{customer } i \text{ gets back his own hat}\} .$$

$$\text{Then } X = X_1 + X_2 + \dots + X_n.$$

Since the ordering of hats is random, each customer has a probability of  $1/n$  of getting back his or her own hat. In other words,  $\Pr\{X_i = 1\} = 1/n$ , which, by Lemma 5.1, implies that  $E[X_i] = 1/n$ .

Thus,

$$\begin{aligned}
 E[X] &= E\left[\sum_{i=1}^n X_i\right] \\
 &= \sum_{i=1}^n E[X_i] \quad (\text{linearity of expectation}) \\
 &= \sum_{i=1}^n 1/n \\
 &= 1,
 \end{aligned}$$

and so we expect that exactly 1 customer gets back his own hat.

Note that this is a situation in which the indicator random variables are *not* independent. For example, if  $n = 2$  and  $X_1 = 1$ , then  $X_2$  must also equal 1. Conversely, if  $n = 2$  and  $X_1 = 0$ , then  $X_2$  must also equal 0. Despite the dependence,  $\Pr\{X_i = 1\} = 1/n$  for all  $i$ , and linearity of expectation holds. Thus, we can use the technique of indicator random variables even in the presence of dependence.

### Solution to Exercise 5.2-5

Let  $X_{ij}$  be an indicator random variable for the event where the pair  $A[i], A[j]$  for  $i < j$  is inverted, i.e.,  $A[i] > A[j]$ . More precisely, we define  $X_{ij} = I\{A[i] > A[j]\}$  for  $1 \leq i < j \leq n$ . We have  $\Pr\{X_{ij} = 1\} = 1/2$ , because given two distinct random numbers, the probability that the first is bigger than the second is  $1/2$ . By Lemma 5.1,  $E[X_{ij}] = 1/2$ .

Let  $X$  be the the random variable denoting the total number of inverted pairs in the array, so that

$$X = \sum_{i=1}^{n-1} \sum_{j=i+1}^n X_{ij}.$$

We want the expected number of inverted pairs, so we take the expectation of both sides of the above equation to obtain

$$E[X] = E\left[\sum_{i=1}^{n-1} \sum_{j=i+1}^n X_{ij}\right].$$

We use linearity of expectation to get

$$\begin{aligned}
 E[X] &= E\left[\sum_{i=1}^{n-1} \sum_{j=i+1}^n X_{ij}\right] \\
 &= \sum_{i=1}^{n-1} \sum_{j=i+1}^n E[X_{ij}] \\
 &= \sum_{i=1}^{n-1} \sum_{j=i+1}^n 1/2
 \end{aligned}$$

$$\begin{aligned}
&= \binom{n}{2} \frac{1}{2} \\
&= \frac{n(n-1)}{2} \cdot \frac{1}{2} \\
&= \frac{n(n-1)}{4}.
\end{aligned}$$

Thus the expected number of inverted pairs is  $n(n-1)/4$ .

### Solution to Exercise 5.3-2

Although PERMUTE-WITHOUT-IDENTITY will not produce the identity permutation, there are other permutations that it fails to produce. For example, consider its operation when  $n = 3$ , when it should be able to produce the  $n! - 1 = 5$  non-identity permutations. The **for** loop iterates for  $i = 1$  and  $i = 2$ . When  $i = 1$ , the call to RANDOM returns one of two possible values (either 2 or 3), and when  $i = 2$ , the call to RANDOM returns just one value (3). Thus, PERMUTE-WITHOUT-IDENTITY can produce only  $2 \cdot 1 = 2$  possible permutations, rather than the 5 that are required.

### Solution to Exercise 5.3-4

PERMUTE-BY-CYCLIC chooses *offset* as a random integer in the range  $1 \leq \text{offset} \leq n$ , and then it performs a cyclic rotation of the array. That is,  $B[(i + \text{offset} - 1) \bmod n + 1] = A[i]$  for  $i = 1, 2, \dots, n$ . (The subtraction and addition of 1 in the index calculation is due to the 1-origin indexing. If we had used 0-origin indexing instead, the index calculation would have simplified to  $B[(i + \text{offset}) \bmod n] = A[i]$  for  $i = 0, 1, \dots, n - 1$ .)

Thus, once *offset* is determined, so is the entire permutation. Since each value of *offset* occurs with probability  $1/n$ , each element  $A[i]$  has a probability of ending up in position  $B[j]$  with probability  $1/n$ .

This procedure does not produce a uniform random permutation, however, since it can produce only  $n$  different permutations. Thus,  $n$  permutations occur with probability  $1/n$ , and the remaining  $n! - n$  permutations occur with probability 0.