## Homework #4

## 1. (10 points) Exercise 5.2-4 (page 122): Hat-check problem.

Ans:

Let the random variable *X* be the number of men that get their own hat. Then, by the definition of expectation, we have:

$$\operatorname{Ex}(X) = \sum_{k=0}^{n} k \cdot \Pr(X = k)$$

However, it is very difficult to compute the probability Pr(X=k) for k=1,2,...,n, because the event for a man gets his own hat is NOT mutually independent. Even so we can solve the problem by using **linearity of expectation**.

Let  $X_i$  be an indicator for whether the *i*-th man gets his own hat or not. That is,  $X_i=1$  means the i-th man gets his own hat, and  $X_i=0$  means the i-th man gets the wrong hat. So the number of men that get their own hat is the sum of all these indicators:

$$X = \sum_{i=0}^{n} X_i$$

Then take the expected value of both sides and apply linearity of expectation:

$$\operatorname{Ex}(X) = \operatorname{Ex}(\sum_{i=0}^{n} X_i) = \sum_{i=0}^{n} \operatorname{Ex}(X_i)$$

We know that every single man is as likely to get one hat as another, so the probability for a man to get his own hat is just 1/n. So, we have:

$$\operatorname{Ex}(X_i) = 1 \cdot \Pr(X_i = 1) + 0 \cdot \Pr(X_i = 0) = \frac{1}{n}$$

Then, 
$$\operatorname{Ex}(X) = \sum_{i=0}^{n} \operatorname{Ex}(X_i) = n \cdot \frac{1}{n} = 1.$$

2. (90 points) Problem 5-2 (page 143-144): Searching an unsorted array. Only: (a), (b), (e), (h) only case k=1, and (i) Assume that the number of indices i such that A[i] = x is 1.

Ans:

A is an unsorted array consisting of n elements, searching for a value x in A.

(a) Pseudocode of random search:

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RANDOM-SEARCH (A, x):
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- 1. n = A.length
- 2. let visited[1..n] be a new bool array
- 3. for i = 1 to n
- 4. visited[i] = FALSE
- 5.  $visited\_counter = 0$
- 6. while visited\_counter < n
- 7. r = RANDOM(1,n) // Choose a random number between 1 and n
- 8. if visited[r]
- 9. continue
- 10. else
- 11. if A[r] == x
- 12. return r
- 13. visited[r] = TRUE
- 14. visited\_counter += 1
- 15. return 0 // FAIL
- (b) Let X be the number of indices into A that we pick before we find x. Then "X=i" means we does not find x for the first (i-1) random choices of index and hit x at the i-th picking. Since there is exactly one index i such that A[i] = x, the probability of hitting x for each random choice is 1/n, where n is the length of A. Therefore, the probability of X=i is  $\frac{1}{n}(1-\frac{1}{n})^{i-1}$ . By the definition of expectation, we have:

$$\operatorname{Ex}(X) = \sum_{i=0}^{\infty} i \cdot \Pr(X = i) = \sum_{i=0}^{\infty} i \cdot \frac{1}{n} (1 - \frac{1}{n})^{i-1}$$

Let 
$$S = \sum_{i=1}^{\infty} i \cdot (1 - \frac{1}{n})^{i-1} = 1 \cdot (1 - \frac{1}{n})^0 + 2 \cdot (1 - \frac{1}{n})^1 + 3 \cdot (1 - \frac{1}{n})^2 + \dots + (i+1) \cdot (1 - \frac{1}{n})^i + \dots$$

Then, 
$$(1-\frac{1}{n})\cdot S = \sum_{i=1}^{\infty} i\cdot (1-\frac{1}{n})^i = 1\cdot (1-\frac{1}{n})^1 + 2\cdot (1-\frac{1}{n})^2 + \dots + i\cdot (1-\frac{1}{n})^i + \dots$$

So, 
$$S - (1 - \frac{1}{n}) \cdot S = \sum_{i=0}^{\infty} (1 - \frac{1}{n})^i = (1 - \frac{1}{n})^0 + (1 - \frac{1}{n})^1 + \dots + (1 - \frac{1}{n})^i + \dots$$

That is 
$$\frac{1}{n} \cdot S = \sum_{i=1}^{\infty} (1 - \frac{1}{n})^i = \frac{1 \cdot (1 - (1 - \frac{1}{n})^{\infty})}{1 - (1 - \frac{1}{n})} = n$$
 where we know that  $S = n^2$ .

Finally, we have:

$$\mathsf{Ex}(X) = \sum_{i=0}^{\infty} i \cdot (1 - \frac{1}{n})^{i} \cdot \frac{1}{n} = \frac{1}{n} \sum_{i=0}^{\infty} i \cdot (1 - \frac{1}{n})^{i} = \frac{1}{n} S = n$$

(e) We know that A[1], A[2], ..., A[i] will be searched before the algorithm terminates if A[i] = x. Because there is exactly one index i such that A[i] = x and all possible permutations of the input array are equally likely, we know that the probability that A[i] = x for i = 1, 2, ..., n, is 1/n. Let N be the number of searched elements before we find x, then we know that N = i if A[i] = x.

Assume that the comparison of each search step costs a constant time c, the average running time will be:

$$T_{avg}(n) = c \cdot \text{Ex}(N) = c \cdot \sum_{i=1}^{n} i \cdot \Pr(A[i] = x) = c \cdot \frac{1}{n} \frac{n(n+1)}{2} = \frac{c}{2}(n+1)$$

The worst case happens when the x appears at the end of A, and the running time is:

$$T_{worst}(n) = c \cdot N(A[n] = x) = cn$$

(h) Assume that the permutation of one element costs  $c_1$  and comparison for search costs  $c_2$ . We only consider case k=1, where there is exactly one index i such that A[i] = x. The running time of SCRAMBLE-SEARCH should be the sum of both permuting and searching cost.

The worst case is that x is the last element of the resulting permuted array. According to (e), we know that:

$$T_{avg}(n) = c_1 \cdot n + c_2 \cdot \operatorname{Ex}(N) = c_1 n + \frac{c_2}{2}(n+1),$$

$$T_{worst}(n) = c_1 \cdot n + c_2 \cdot n$$

(i) By the assumption that the index i = 1 for A[i] = x. Because A[1] = x, we would choose DETERMINISTIC-SEARCH so that the algorithm will find x and terminate by picking only one element.