MSP 2024 - Tutorial 3

Exercise 1. Consider a stream of n elements where n is not known in advance.

- **1** Write a program in $\mathcal{O}(1)$ space that returns a random stream element, where all elements have equal probability of being picked.
- Show correctness.

2.

3.

stream_random_element(stream):
1. item = stream.next()

while not stream.empty():

 $item_count = 1$

```
4. next = stream.next()
5. item_count += 1
6. r = random_float(0,1)
7. if r < 1/item_count:
8. item = next
9. return item

Correctness:
P(\text{i-th element is returned}) = P(\text{i-th element is stored to variable item}) \cdot P(\text{item is not rewritten afterwards}) = \frac{1}{i} \cdot \left(\frac{i}{i+1} \cdot \frac{i+1}{i+2} \cdots \frac{n-2}{n-1} \cdot \frac{n-1}{n}\right) = \frac{1}{n}
```

Exercise 2. Assume an array with n elements.

- ① Write a program that generates a random permutation of this array in $\mathcal{O}(n)$ time.
- 2 Show correctness.

Exercise 3. Assume the following modification of the hiring problem, where the company must pay severance to the fired worker if they have just been hired:

```
Hire-Assistant(n)
1. hire 1, best = 1
2. for i from 2 to n:
3.    if candidate i is better than best:
4.      fire best
5.      if best == i-1:
6.      pay severance to best
7.    hire i, best = i
```

Assume that candidates arrive at the interview in random order. Compute how many times the severance will be paid:

a) in the best case, b) in the worst case, c) on average.

- a 0 times, e.g. if candidate 1 is the best one
- **b** n-1 times iff the candidates are sorted from worst to best
- **c** Let X_i , $2 \le i \le n$, be the indicator variable that attains value 1 if the severance is paid during i-th iteration, and 0 otherwise. The severance is paid during i-th iteration when (i-1)-th as well as i-th candidates were hired, that is, if candidate #(i-1) is the best among first (i-1) candidates and candidate #i is the best among first i candidates. Thus, $P(X_i = 1) = \frac{1}{i-1} \cdot \frac{1}{i}$.

The expected total number of paid severances is then

$$E\left[\sum_{i=2}^{n} X_i\right] = \sum_{i=2}^{n} E\left[X_i\right] = \sum_{i=2}^{n} P(X_i = 1) = \sum_{i=2}^{n} \frac{1}{i-1} \cdot \frac{1}{i} = \sum_{i=2}^{n} \frac{1}{i-1} - \frac{1}{i} = \left(1 - \frac{1}{2}\right) + \left(\frac{1}{2} - \frac{1}{3}\right) + \dots + \left(\frac{1}{n-2} - \frac{1}{n-1}\right) + \left(\frac{1}{n-1} - \frac{1}{n}\right) = 1 - 1/n$$

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Exercise 4. Consider the following algorithm describing one pass of the *bubble-sort* algorithm. Assume that the input array a (indexed from 1) contains numbers from 1 to n in random order (all permutations have equal probability).

```
bubble-sort-one-pass(a,n):
```

- 1. for i from 1 to n-1:
- 2. if a[i] > a[i+1]:
- 3. swap a[i] and a[i+1]

Determine:

- Probability that no elements are swapped (best-case behaviour).
- **2** Probability that n-1 swaps are executed (worst-case behaviour).
- 3 The expected number of swaps (average-case behaviour) it is sufficient to give the asymptotic number of swaps including an explanation.

- **1** No elements are swapped when array is sorted: $P = \frac{1}{n!}$
- 2 n-1 swaps are executed when a [1] contains n: $P = \frac{1}{n}$ 3 Let X_i , $1 \le i \le n-1$, be the indicator variable that attains value 1 if a
- swap is performed during i-th iteration, and 0 otherwise. Right before the i-th iteration, a[i] contains the maximum value between a[1]...a[i]. Then, X_i is 0 when a[i+1] is the largest number between a[1]...a[i+1], i.e. with probability $\frac{1}{i+1}$. Thus, $P(X_i = 1) = 1 \frac{1}{i+1}$.

The expected total number of swaps is then

$$E\left[\sum_{i=1}^{n-1} X_i\right] = \sum_{i=1}^{n-1} E\left[X_i\right] = \sum_{i=1}^{n-1} P(X_i = 1) = \sum_{i=1}^{n-1} 1 - \frac{1}{i+1} =$$

$$= \sum_{i=1}^{n-1} 1 - \sum_{i=1}^{n-1} \frac{1}{i+1} = n - 1 - \sum_{i=1}^{n-1} \frac{1}{i+1} =$$

$$= \mathcal{O}(n) - \mathcal{O}(\log n) = \mathcal{O}(n)$$

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Exercise 5. Consider the following randomised algorithm all_even that tests whether array arr of size len > 0 (indexed from 1) contains only even numbers. Function rand_int(1,len) returns a random integer in range 1 to len with the uniform probability.

```
all_even(arr,len,n):
1. for i from 1 to n:
2.     k = rand_int(1,len)
3.     if arr[k] is odd:
4.         return false
5. return true
```

- Decide and justify whether all_even is a Las Vegas or a Monte Carlo algorithm.
- ② Construct function $\mathcal{P}(\text{len},n)$ that returns, for the given len and n>0, the upper bound on the probability (with respect to all input arrays of size len), that all_even(arr,len,n) returns a wrong result.
- 3 Determine the smallest n such that the worst-case probability that all_even(arr,10,n) returns a wrong result is smaller than 50%.

- 1 It is a Monte Carlo algorithm because there is a non-zero probability that for a given input the algorithm returns a wrong result.
- **2** Clearly, if len = 1, then $\mathcal{P}(\text{len}, n) = 0$. It is easy to see that the worst-case input contains only a single odd number. For such inputs $\mathcal{P}(\text{len}, n) = \left(\frac{len-1}{len}\right)^n$.
- 3 Since $P(10, n) = \left(\frac{9}{10}\right)^n$, we require that $\left(\frac{9}{10}\right)^n < 0.5$, i.e. $\log_{0.9}(0.9^n) > \log_{0.9}(0.5) \Rightarrow n > 6.5$. Hence n = 7.