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1. (a) *Use the subroutine `probefluxcapacitor()` to implement the routine `randbit()` that outputs a uniformly random bit.*

Pseudocode

```
def randbit():  
    while true:  
        (bitA, p) = probefluxcapacitor()  
        (bitB, p) = probefluxcapacitor()  
        if bitA does not equal bitB:  
            return bitA
```

Correctness

If the bits were uniformly random then a 1 or a 0 would be equally probable. This would mean that the probability that either one would occur would need to be $\frac{1}{2}$ in order to be uniformly random.

For the problem we are given that $P(1) = p$ and $P(0) = (1 - p)$. The first thing we have to figure out is how do we make an algorithm such that the probability of returning a 1 is equal to the probability of returning a 0. The second thing we have to figure out is how do we make sure the final probability for any run of `randbit()` is equal $\frac{1}{2}$.

Notice the probability of two independent events, A and B , occurring together is $P(A \cap B) = P(A) \cdot P(B)$ [1]. If we let those two events be the probability of `randbit()` returning a 1, $P(1)$, and the probability of `randbit()` returning a 0, $P(0)$, then we can say that the probability of them both occurring, $P(1 \cap 0)$, is $p \cdot (1 - p)$. This helps because we know that one call to `probefluxcapacitor()` will either return p , for bit value 1, or $(1 - p)$, for bit value 0. But if we combine two

calls to `probefluxcapacitor()` sequentially and then compare the results we can create a probability like $P(1 \cap 0)$.

Two calls to `probefluxcapacitor()` could result in 4 four combinations:

- $P(1 \cap 1) = p \cdot p$
- $P(0 \cap 0) = (1 - p) \cdot (1 - p)$
- $P(1 \cap 0) = p \cdot (1 - p)$
- $P(0 \cap 1) = (1 - p) \cdot p$

The pseudocode above only returns a bit if the probability combination was $p \cdot (1 - p)$ or $(1 - p) \cdot p$ which are in fact equal to each other. This tells us that returning a 1 — bitA is 1 and bitB is 0 — and returning a 0 — bitA is 0 and bitB is 1 — have equal probability now. And since the if statement only looks at two possible combinations from the two calls to `probefluxcapacitor()` and out of only one of those possible combinations will a 1 be returned, the probability of returning a 1, $P(1)$, is $\frac{1}{2}$.

Runtime

In order to find the running time of a randomize algorithm, we need to look at the expectation of the running time. The two calls to `probefluxcapacitor()` each take $O(1)$ because they are just looking at at most two numbers. The checking of the two bits in the if statement is also $O(1)$. This means that one time through the loop takes $O(1)$. The probability of getting out of the while loop and returning is $p \cdot (1 - p)$ for a 1 or $(1 - p) \cdot p$ for a 0.

Let X be the indicator random variable associated with the returning of `rand-bit()`. X_1 means a 1 is returned and X_0 means a 0 is returned. This also means $E[X_1] = P(1)$, the probability of 1 being returned, and $E[X_0] = P(0)$, the prob-

ability of 0 being returned. This gives us

$$\begin{aligned} E[X] &= \sum_{i=0}^1 E[X_i] \\ &= E[X_0] + E[X_1] \\ &= p \cdot (1 - p) + (1 - p) \cdot p \\ &= 2p(1 - p) \end{aligned}$$

This means the running time of `randbit()` is $O(2p(1 - p) \cdot 1) = O(2p(1 - p))$.

- (b) *Implement the algorithm randbit(p) that outputs an independent random bit with $P(\text{randbit}(p)=1)=p$.*

Pseudocode

```
def randbit(p):  
    l = list of bits  
    for i from 1 to n:  
        bit = randbit()  
        append bit to l  
    if 1 in l:  
        return 1  
    else:  
        return 0
```

Correctness

randbit() will return a 1 with a probability p of $\frac{1}{2}$. We consider putting the independent event A of running randbit() a single time into a collection. Then running randbit() n times would give us a collection of n events/bits. Since these events are all independent then the probability of them happening together is

$$\begin{aligned} &= P(A_1) \cdot P(A_2) \cdot \dots \cdot P(A_n) \\ &= \frac{1}{2} \cdot \frac{1}{2} \cdot \dots \cdot \frac{1}{2} \\ &= \frac{1}{2^n} \end{aligned}$$

k is the number of outcomes we want out of the combined of the runs of randbit(). For example, if we want $P(\text{randbit}(\frac{3}{4}) = 1) = \frac{3}{4}$, then we would need look at the combined output of calling randbit() two times because $n = 2$. This would give us four possible possible outcomes: (0,0), (1,0), (0,1), (1,1). Out of those four possible outcomes there is a 1 contained in three of them. This means that if we return a 1 when a 1 is in a possible outcome we get $p = 3/4$.

In the pseudocode above we generalize this idea by running randbit() n times and keeping a list of the bits that are returned by randbit(). If a 1 exists in the

list then we return a 1 otherwise we return a 0.

This doesn't appear to work for any other example so it is not correct. I was unable to figure out a solution to this problem.

(c) *Show the correctness and runtime for randbit(p) when p is not of the form $k/2^n$*

Correctness

Let's look at the algorithm when $p = 3/4$. This means $p = 0.11$ and i will be in the range 0 to 1. d is first set to b_0 which is a 1. Then randbit() is called and its output is compared to d . The probability of randbit() returning a 0 which would end the function is $1/2$. The next time through d is set to b_1 which is a 1. Again randbit() is called and its output is compared to d and the function returns if the output of randbit() is not the same as d . Each time, the probability of the output of randbit() not being equal to d is $1/2$.

I don't have any answer for proving the correctness of this problem.

Runtime

We are given that getdigit(p, i) runs in $O(i^c)$ time for some c . The if statement where the output of randbit() is compared to d has a probability of $1/2$ of being correct each time through the loop. The loop itself could run a maximum total of i times which is equal to the number of digits in the binary representation of p .

The upper bound on this function is $O(i(i^c + 1/2))$

References

- [1] <https://www.mathsisfun.com/data/probability-events-independent.html>
- [2] <https://www.mathsisfun.com/data/binomial-distribution.html>