- 1. (a) Each time you roll a die, the chance of getting a six is p = 1/6. So the expected number of rolls until you see a six is 1/p = 6.
 - (b) Each time you pull out a random fish, the chance of getting a catfish is p = 50/1000 = 1/20. Thus the expected number of fish you need to pull out until you get a catfish is 1/p = 20.
 - (c) Each time the computer chooses a random integer, it has a probability p = 1/10 of getting a multiple of 10. Therefore the expected number of trials is 1/p = 10.
- 2. (a) On any iteration, the probability of outputting H is p(1-p) (probability of getting heads, then tails) and the probability of outputting T is (1-p)p (probability of getting tails, then heads). Since these are equal, both outcomes are equally likely.
 - (b) The probability that a particular iteration is successful (that is, the algorithm halts) is 2p(1-p). Thus the expected number of iterations is 1/(2p(1-p)) and the expected number of coin tosses is twice this, 1/(p(1-p)).
- 3. Here's the algorithm, given input x:

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Repeat 100 times: Run \mathcal{A}(x) If it says ''not prime'', output ''not prime'' and halt Output ''prime''
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If x is "prime", then \mathcal{A} will always return "prime", and hence the answer will be correct. If x is not prime, then the probability that $\mathcal{A}(x)$ returns "prime" is at most 1/2, and the probability that it returns "prime" 100 times is at most $1/2^{100}$.

4. (a) Here's an algorithm that makes use of both \mathcal{A} and \mathcal{B} . On input x,

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Run \mathcal{A}(x) If it says ''not prime'': output ''not prime'' and halt Run \mathcal{B}(x) and output the answer
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To see why this works, suppose first that x is prime. Then $\mathcal{A}(x)$ will return "prime" and thus $\mathcal{B}(x)$ will be invoked, returning the right answer. On the other hand, if x is not prime, then either $\mathcal{A}(x)$ will detect this, or $\mathcal{B}(x)$ will be invoked.

- (b) If the input is prime, both procedures will be called, for a running time of 101T(n).
- (c) If the input is not prime, then $\mathcal{A}(x)$ will detect this with probability at least 1/2; otherwise $\mathcal{B}(x)$ will also need to be called. Thus the expected running time is $T(n) + 0.5 \times 100T(n) = 51T(n)$.
- 5. This is equivalent to throwing n balls in n bins. The size of the largest bin is (with high probability) $O(\log n)$, and this is therefore the number of hours the repairs would take overall.
- 6. We saw in class that when n balls are thrown into n^2 bins, there is at least a 1/2 probability that there will be no collisions. Therefore, we should set $2^m = n^2$, roughly, which means $m = 2 \log n$.
- 7. Hashing with open addressing.
 - (a) If n items are stored in a table of size 2n, half the table is empty. Let's say we are inserting x. The locations $h(x,0), h(x,1), \ldots$ are random and thus each of them has at least a 1/2 chance of being empty. The probability the first k locations are all occupied is therefore at most $1/2^k$.
 - (b) Let A_i be the event that the *i*th insertion requires more than k probes.

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\Pr(\text{some insertion requires more } k \text{ probes}) = \Pr(A_1 \cup A_2 \cup \dots \cup A_n) \leq \Pr(A_1) + \dots + \Pr(A_n) \leq \frac{n}{2^k}
For k = 2\log_2 n, this is 1/n.
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- 8. Skip lists.
 - (a) Roughly speaking, a random element will on average lie halfway down the list and will thus need about n/2 time to locate. More precisely, the time taken to find the *i*th item in the list is *i*. For an element chosen at random, the expected lookup time is thus

$$\sum_{i=1}^{n} \Pr(\text{item } i \text{ is chosen}) \ i = \frac{1+2+\dots+n}{n} = \frac{n+1}{2}.$$

The worst-case lookup time is n.

- (b) Very roughly, a random element will on average be about halfway down the list, and will therefore be reached by about n/(2k) jump pointers and k/2 next pointers, for a total time of n/(2k) + k/2. The worst-case lookup time is n/k + k.
- (c) The best choice is $k \approx \sqrt{n}$, leading to an expected and worst-case lookup time of $O(\sqrt{n})$.
- 9. We saw in class that when n items are stored in a Bloom filter of size m, using k hash functions, the fraction of zero entries in the table is roughly $e^{-kn/m}$. Therefore, a reasonable way to estimate n is as follows:
 - Determine the fraction of zero entries in the table; call this q.
 - Return $(m/k) \ln(1/q)$.
- 10. The secretary problem.
 - (a) Let s_1 denote the best secretary and s_2 the second-best secretary. If s_2 is one of the first r candidates (which happens with probability r/n) and s_1 is one of the remaining n-r candidates (which happens with probability (n-r)/n), then Barbara will correctly identify s_1 .

$$\Pr(\text{Barbara chooses } s_1) \ge \frac{r}{n} \cdot \frac{n-r}{n} = \frac{r(n-r)}{n^2}.$$

(a) Setting r = n/2 results in a 1/4 probability of success.