Probability and Computing, 2nd Edition

Solutions to Chapter 5: Balls, Bins, and Random Graphs

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5.1

As $(1+1/n)^n$ increases, we find the smallest n to reach the threshold. $(1+1/n)^n$ first reaches 0.99e at n=50, and 0.999999e at n=499982. Since $(1-1/n)^n$ also increases, we solve in a similar way. $(1-1/n)^n$ first reaches 0.99/e at n=51 and 0.999999/e at n=499991.

5.2

Recall the formula used in the birthday paradox: If there are N possibilities, then we solve for the smallest n that satisfies $\prod_{i=1}^{n-1} (1-\frac{i}{N}) \approx \prod_{i=1}^{n-1} e^{-i/n} = e^{-(n-1)n/2N} < 1/2$. Note that we omitted the final approximation to derive exact numerical answers. Regardless of whether the number of Social Security number digits is 9 or 13, using the last four digits gives N = 10000 and this gives n = 119. In the case where the number of digits is 9 $(N = 10^9)$, we get n = 37234. In the case where the number of digits is 13 $(N = 10^{13})$, we get n = 3723298.

5.3

Let the number of balls thrown be m. Then the desired probability is $\prod_{i=0}^{m-1} (1-\frac{i}{n})$. We first determine c_1 . $m=c_1\sqrt{n}$ should satisfy $\prod_{i=0}^{m-1} (1-\frac{i}{n}) \leq \prod_{i=0}^{m-1} e^{-i/n} = e^{-(m-1)m/2n} \leq e^{-1}$. Since $(m-1)m=c_1^2n-c_1\sqrt{n} \geq 2n$, $(c_1^2-2)\sqrt{n} \geq c_1$. Therefore, we choose c_1 that is greater than or equal to $\frac{1}{2}\left(\frac{1}{\sqrt{n}}+\sqrt{\frac{1}{n}}+8\right)$. Now we determine c_2 . To use the given hint, assume that 2m < n. $\prod_{i=0}^{m-1} (1-\frac{i}{n}) \geq \prod_{i=0}^{m-1} \exp(-\frac{i}{n}-\frac{i^2}{n^2}) = \exp(-\frac{m(m-1)}{2n}-\frac{(m-1)m(2m-1)}{6n^2}) = \exp(-\frac{m(m-1)}{2n}(1+\frac{2m-1}{3n})) \geq \exp(-\frac{m^2}{2n}(1+\frac{2m}{3\sqrt{n}})) \geq \frac{1}{2}$ should be satisfied for $m=c_2\sqrt{n}$. This is equivalent to satisfying $\frac{c_2^2}{2}(1+\frac{2c_2}{3\sqrt{n}}) \leq \ln 2$. Since n is sufficiently large, choosing $c_2=\sqrt{2\ln 2-\frac{1}{\ln n}}$ yields the desired result.

5.4

Let event A indicate that there exist two or more people who share a birthday, and event B indicate that exactly two people share a birthday. Then our desired probability would be $\Pr(A - B) = \Pr(A) - \Pr(B)$ since $B \subset A$.

We first determine Pr(A), which is easy: $Pr(A) = 1 - \prod_{i=1}^{100} \frac{366-i}{365}$.

We now determine Pr(B). If there are i shared birthdays in which each day is

shared by exactly two people, then the number of possible permutations would be the product of the following terms:

 $\binom{365}{i}$ ways to choose *i* shared days, $\binom{100}{2i}$ ways to choose 2i people to share birthdays, $\prod_{i=1}^{i} {2j \choose 2}$ ways to distribute *i* birthdays to 2i people and $\prod_{i=1}^{100-2i} (366 - 1)^{-2i}$ (i-j) ways to distribute unique birthdays to the rest.

Thus,
$$\Pr(B) = \sum_{i=0}^{50} \frac{365!100!}{i!(100-2i)!(265+i)!2^i} \times \frac{1}{365^{100}}$$
.

Therefore, we can determine our desired probability Pr(A-B) = Pr(A) - Pr(B).

5.5

Let $X \sim Poisson(\lambda)$. Then $M_X(t) = \mathbf{E}[e^{tX}] = e^{\lambda(e^t-1)}$ holds. By computing the second derivative of $M_X(t)$ with respect to t and plugging t=0 in, we get $\mathbf{E}[X^2] = \lambda + \lambda^2$. Thus, $\mathbf{Var}[X] = \lambda$ follows.

5.6

We first show that $Y \sim Poisson(\mu p)$.

$$\Pr(Y = k) = \sum_{i=k}^{\infty} \Pr(X = i) {i \choose k} p^k (1-p)^{i-k} = \sum_{i=k}^{\infty} \frac{e^{-\mu} \mu^i}{i!} \frac{i!}{k!(i-k)!} p^k (1-p)^{i-k}$$
$$= \frac{e^{-\mu} (p\mu)^k}{k!} \sum_{i=k}^{\infty} \frac{(\mu(1-p))^{i-k}}{(i-k)!} = \frac{e^{-\mu} (p\mu)^k}{k!} e^{\mu(1-p)} = \frac{e^{-\mu p} (\mu p)^k}{k!}.$$

We can also similarly show that $Z \sim Poisson(\mu(1-p))$.

Now we show that Pr(Y = i, Z = j) = Pr(Y = i) Pr(Z = j). Note that X = Y + Z by definition. This allows us to write Pr(Y = i, Z = j) as Pr(Y = i, Z = j) $i, X = i + j) = \Pr(X = i + j) {i+j \choose i} p^i (1-p)^j = \frac{e^{-\mu} \mu^{i+j}}{i!j!} p^i (1-p)^j.$ Since $\Pr(Y = i) \Pr(Z = j) = \frac{e^{-p\mu} (p\mu)^i}{i!} \frac{e^{-(1-p)\mu} ((1-p)\mu)^j}{j!} = \frac{e^{-\mu} \mu^{i+j} p^i (1-p)^j}{i!j!} = \Pr(X = i + j) \frac{e^{-p\mu} (p\mu)^i}{i!} \frac{e^{-(1-p)\mu} ((1-p)\mu)^j}{j!} = \frac{e^{-\mu} \mu^{i+j} p^i (1-p)^j}{i!j!} = \frac{e^{-\mu} \mu^{i+j} p^i (1-p)^j}{i!j!}$

Since
$$\Pr(Y = i) \Pr(Z = j) = \frac{e^{-p\mu}(p\mu)^i}{i!} \frac{e^{-(1-p)\mu}((1-p)\mu)^j}{j!} = \frac{e^{-\mu}\mu^{i+j}p^i(1-p)^j}{i!j!} = \Pr(Y = i, X = i+j), Y \perp\!\!\!\perp Z. \blacksquare$$

5.7

We first prove that $\ln(1+x) \le x$, which is equivalent to $1+x \le e^x$. Since $\ln(1+x) - x = -\frac{x^2}{2} + \frac{x^3}{3} - \cdots$, this can be seen as an alternating series as $\frac{x^n}{n}$ is monotonically decreasing in $|x| \le 1$. We can apply rearrangements to the alternating series as $\ln(1+x) - x = -\sum_{n=1}^{\infty} \left(\frac{x^{2n}}{2n} - \frac{x^{2n+1}}{2n+1}\right)$, since the Taylor expansion of $\ln(1+x)$ is absolutely convergent (to e^x-1). The rearrangement gives $\ln(1+x) - x \le 0$, which is the desired result.

We now prove $x + \ln(1-x^2) \le \ln(1+x)$, which is equivalent to $e^x(1-x^2) \le 1+x$. Since $\ln(1-x^2) = \ln(1+x) + \ln(1-x)$, $x + \ln(1-x^2) \le \ln(1+x)$ is reduced to $\ln(1-x) \le -x$. At $|x| \le 1$, this is equivalent to $\ln(1+x) \le x$, which we have previously proved.