STAT 598G Spring 2011 Jeremiah Rounds and Sergey Kirshner Homework 2 (Written Exercises)

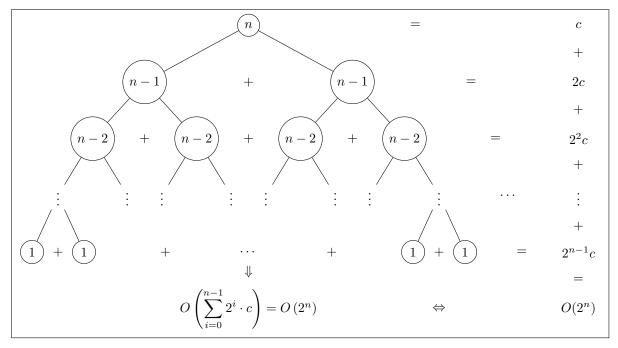
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1. (2 pts) http://learning.stat.purdue.edu/wiki/_media/courses/sp2011/598g/intro.tar.gz contains file recursion.c which did not examine in class. What is the recurrence that is being solved by this program? What is the order of the running time complexity for the function f(n)?

Solution: The program is computing the recurrence f(n) = f(n-1) + f(n-2) for $n \in \mathbb{N}$. The base case for the recurrence is f(1) = 1 and f(0) = 0. The second part is tricky. Denote by T(n) the number of operations needed to compute f(n). Checking for n > 1 and adding two numbers takes constant time, say total of c. (We can assume that returning n if $n \le 1$ takes the same time c.) Then T(n) = T(n-1) + T(n-2) + c, the same type of recurrence f is solving!

Instead of finding the exact order Θ , we will instead find the asymptotic upper bound O. Note that $T(n) \ge T(n-1)$ for $n \ge 1$. Therefore

$$T(n) = T(n-1) + T(n-2) + c \le 2T(n-1) + c.$$



The illustration in the figure above suggests that $T(n) \in O(2^n)$. (Enough for full credit.) 2^n is however a loose asymptotic upper bound. One can actually show that $T(n) \in \Theta\left(\left(\frac{\sqrt{5}+1}{2}\right)^n\right)$ (bonus part of programming homework 2, exercise 1).

2. (1 pts) Exercise 2.1 (Robert and Casella, 2009): For an arbitrary random variable X with cdf F, define the generalized inverse of F by

$$F^-\left(u\right)=\inf\left\{x:\ F(x)\geq u\right\}.$$

Show that if $U \sim \text{Unif}(0,1)$, then $F^{-}(U)$ is distributed like X.

Solution: Let $Y = F^{-}(U)$. Then

$$P(Y \le y) = P(F^{-}(U) \le y).$$

Assume $u \in (0,1)$. For any (y,u), $F^-(u) \leq y$ implies $F(y) \geq u$ (from the definition of $F^-(u)$). On the other hand, for any (y,u) such that $F(y) \geq u$, $F^-(u) \leq y$ (again from the definition of $F^-(u)$). Thus $\{(y,u): F^-(u) \leq y\} = \{(y,u): F(y) \geq u\}$. So

$$P\left(Y\leq y\right)=P\left(F^{-}\left(U\right)\leq y\right)=P\left(F\left(y\right)\geq U\right)=P\left(U\leq F\left(y\right)\right)=F\left(y\right)=P\left(X\leq y\right).$$

3. (1 pts) Gumbel distribution has the cumulative distribution function

$$F(x; \mu, \beta) = e^{-e^{-(x-\mu)/\beta}}, \ x \in \mathbb{R}.$$

Assuming one can draw samples from $\mathcal{U}(0,1)$, describe (and prove) how to draw samples from F.

Solution: We apply the inverse transformation method to sample from F. To do so, we note that F is continuous and is therefore invertible on its range of (0,1). Now, assume $U \sim \mathsf{Unif}(0,1)$. Now, we compute the inverse transformation

$$u = F\left(x\right) = e^{-e^{-(x-\mu)/\beta}} \Leftrightarrow -\ln u = e^{-(x-\mu)/\beta} \Leftrightarrow -\ln \left(-\ln u\right) = \frac{x-\mu}{\beta} \Leftrightarrow x = \mu - \beta \ln \left(-\ln u\right).$$

Thus $F^{-1}(u) = \mu - \beta \ln(-\ln u)$ for $u \in (0,1)$. One can obtain samples $x \sim F$ by drawing $u \sim \text{Unif } (0,1)$, and then setting $x = \mu - \beta \ln(-\ln u)$.

4. (1 pts) Prove that the Box-Muller procedure produces two independent standard normal random variables.

Solution: We pick up where the notes left off. Assuming $u_1, u_2 \stackrel{iid}{\sim} \text{Unif } (0,1), f_{U_1U_2}(u_1, u_2) = 1$ if $u \in (0,1), v \in (0,1)$ and 0 otherwise,

$$f_{Z_1Z_2}(z_1, z_2) = f_{U_1U_2}(h_1(z_1, z_2), h_2(z_1, z_2)) |\det J(z_1, z_2)| = \frac{1}{2\pi} \exp\left(-\frac{1}{2}(z_1^2 + z_2^2)\right)$$

for z_1, z_2 s.t. $h_1(z_1, z_2) \in (0, 1)$, $h_2(z_1, z_2) \in (0, 1)$ which it is for $(z_1, z_2) \in \mathbb{R}^2$. Computing marginals of $f_{Z_1Z_2}(z_1, z_2)$ yields $Z_1, Z_2 \sim \mathcal{N}(0, 1)$. Finally, we need to notice that $f_{Z_1}(z_1) f_{Z_2}(z_2) = f_{Z_1Z_2}(z_1, z_2)$ for all $(z_1, z_2) \in \mathbb{R}^2$, so Z_1 and Z_2 are independent.

5. (2 pts) Describe an Accept-Reject sampling algorithm for Rayleigh random variables, and find its probability of acceptance. Rayleigh's pdf is

$$f(x;\sigma) = \frac{x}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}}, \ x \ge 0$$

with a parameter $\sigma > 0$.

Solution: There are several possible solution. Here, we will use $g(x) = e^{-x}$, $x \ge 0$, the density for the standard normal random variable as the proposal distribution for Accept-Reject sampler. Then $r(x) = f(x)/g(x) = \frac{x}{\sigma^2}e^{x-\frac{x^2}{2\sigma^2}}$, $x \ge 0$.

$$h'\left(x\right) = \frac{1}{\sigma^{2}}e^{x - \frac{x^{2}}{2\sigma^{2}}} + \frac{x}{\sigma^{2}}e^{x - \frac{x^{2}}{2\sigma^{2}}}\left(1 - \frac{x}{\sigma^{2}}\right) = -\frac{1}{\sigma^{4}}e^{x - \frac{x^{2}}{2\sigma^{2}}}\left(x^{2} - \sigma^{2}x - \sigma^{2}\right).$$

By solving $x^2 - \sigma^2 x - \sigma^2 = 0$, we find that $h\left(x\right)$ is maximized when $x = \frac{\sigma^2 + \sqrt{\sigma^4 + 4\sigma^2}}{2}$. Denote this value by $\hat{x}\left(\sigma^2\right)$. Then $M\left(\sigma^2\right) = \sup_{x \geq 0} \frac{f(x)}{g(x)} = \frac{\hat{x}\left(\sigma^2\right)}{\sigma^2} \exp\left(-\hat{x}\left(\sigma^2\right) - \frac{1}{2\sigma^2}\left(\hat{x}\left(\sigma^2\right)\right)^2\right)$. One would repeatedly sample $x \sim \mathsf{Exp}\left(1\right)$ and independently $u \sim \mathsf{Unif}\left(0,1\right)$ until $uM\left(\sigma^2\right) \leq r\left(x\right)$, and then accept x in that case. The probability of acceptance of a random sample x is then $1/M\left(\sigma^2\right)$.

6. (1 pts) von Mises distribution is a continuous probability distribution on a circle with pdf

$$f(x|\mu,\kappa) \propto e^{\kappa \cos(x-\mu)}, \ x \in [-\pi,\pi)$$

with parameters $\mu \in \mathbb{R}$ and $\kappa > 0$. Propose an algorithms to draw samples according to f.

Solution: This problem also has multiple solutions. We solve it using Accept-Reject method. Set $\hat{f}(x) = e^{\kappa \cos(x-\mu)}$. Note that $\hat{f}(x) \leq e^{\kappa}$ for all $x \in [-\pi, \pi)$. We will use uniform on $[-\pi, \pi)$, $g(x) = \frac{1}{2\pi}$ as the proposal density. Then

$$\frac{f(x)}{g(x)} = \frac{2\pi \hat{f}(x)}{C} \le \frac{2\pi e^{\kappa}}{C},$$

so we can set $MC = 2\pi e^{\kappa}$ (ensuring that $\frac{f(x)}{g(x)} \leq M$). To draw samples from f, we repeatedly draw $x \sim g = \text{Unif}\left[-\pi, \pi\right)$ and (independently) $u \sim \text{Unif}\left(0, 1\right)$ until $2\pi e^{\kappa} u \leq \frac{\hat{f}(x)}{g(x)}$ (equivalent to $\frac{f(x)}{Mg(x)} \leq u$) upon which we accept x and stop.

7. (2 pts) Gumbel's bivariate exponential distribution has cdf

$$F(x, y; k) = 1 - e^{-x} - e^{-y} + e^{-x - y - kxy}, x, y > 0$$

where $k \in [0,1]$ is a parameter. Describe how to obtain bivariate samples from F.

Solution: We will sample $(x,y) \sim F$ by first sampling $x \sim F_X(x)$, and then $y|x \sim F_{Y|X}(y|x)$. Note that $F_X(x) = \lim_{y \to \infty} F(x,y) = 1 - e^{-x}$; thus both marginals are univariate standard exponentials. Thus one can draw $x \sim \mathsf{Exp}(1)$. For absolutely continuous random variables x and y,

$$F_{y|x}(y|x) = \frac{1}{f_X(x)} \frac{\partial F(x,y)}{\partial x}.$$

Thus

$$F_{Y|X}(y|x) = e^x \left(e^{-x} - (1+ky)e^{-x-y-kxy}\right) = 1 - (1+ky)e^{-(1+kx)y}, \ y > 0.$$

While $F_{Y|X}(y|x)$ cannot be inverted analytically, one can find y corresponding to $u \sim \text{Unif}(0,1)$ using numerical methods (e.g., bisection).