

## 1 Probability.

a.  $X, Y$  i.i.d. r.v.  $X \perp Y$ .  $f_X = f_Y = f$ Let  $Z = \max(X, Y)$ . Prove  $f_Z(x) = 2f(x)P(X \leq x)$ 

To find pdf of  $Z$ , we can find cdf of  $Z$  first and then take the derivative of cdf to find pdf of  $Z$ .

$$\begin{aligned} F_Z(x) &= F_{\max(X, Y)}(x) \\ &= P(\max(X, Y) \leq x) \\ &= P(X \leq x, Y \leq x) \end{aligned}$$

Since  $X$  and  $Y$  are independent,

$$\begin{aligned} P(X \leq x, Y \leq x) &= P(X \leq x) \cdot P(Y \leq x) \\ &= F_X(x) \cdot F_Y(x) \end{aligned}$$

Then, take the derivative of  $F_Z(x)$ .

$$\begin{aligned} f_Z(x) &= \frac{d}{dx}(F_Z(x)) = \frac{d}{dx}(F_X(x) \cdot F_Y(x)) \\ &= F_X(x) \cdot f_Y(x) + f_X(x) \cdot F_Y(x) \\ &= P(X \leq x) \cdot f(x) + f(x) \cdot P(Y \leq x) \end{aligned}$$

Since  $X$  and  $Y$  have same common density function, the cumulative density function for both variables are the same. Thus,  $F_X(x) = F_Y(x)$

$$P(X \leq x) = P(Y \leq x)$$

Therefore, the above calculations can be formalized by

$$\begin{aligned} f_Z(x) &= P(X \leq x)f(x) + f(x)P(X \leq x) \\ &= 2f(x)P(X \leq x) \end{aligned}$$

Q.E.D.

b.  $U \sim \text{unif}(0,1)$ , find dist'n of  $\lfloor 100U \rfloor + 1$ .

$$E(U) = \frac{1}{2}(0+1) = \frac{1}{2}$$

$$\text{Var}(U) = \frac{1}{12}(0-1)^2 = \frac{1}{12}$$

Since  $U$  is a uniform random variable in a closed interval  $[0,1]$ ,  $\lfloor 100U \rfloor$  maps the interval  $[0,1]$  to  $[0,100]$  as when  $u=0$ ,  $\lfloor 100u \rfloor = \lfloor 0 \rfloor = 0$ ; and when  $u=1$ ,  $\lfloor 100u \rfloor = \lfloor 100 \rfloor = 100$ .

Adding 1 to the result maps  $[0,100]$  to  $[1,101]$ . Therefore, the distribution of  $\lfloor 100U \rfloor + 1$  is a discrete uniform distribution in the interval  $[1,101]$ .

Based on properties of uniform distribution,

$$E(\lfloor 100U \rfloor + 1) = \frac{1}{2}(1+101) = 51$$

$$\text{Var}(\lfloor 100U \rfloor + 1) = \frac{1}{12}(1-101)^2 = \frac{10000}{12} = \frac{2500}{3}.$$

c.  $U \sim \text{unif}(0,1)$ ,  $0 < q < 1$ ,  $X = 1 + \lfloor \frac{\log U}{\log q} \rfloor$ .

Prove  $X \sim \text{Geo}(p)$ , and find  $p$ .

$$E(U) = \frac{1}{2}(0+1) = \frac{1}{2}$$

$$\text{Var}(U) = \frac{1}{12}(0-1)^2 = \frac{1}{12}$$

$$F_U(x) = P(U \leq x) = \begin{cases} 0 & \text{for } x < 0 \\ x & \text{for } 0 \leq x \leq 1 \\ 1 & \text{for } x > 1 \end{cases}$$

$$\begin{aligned} F_{\frac{\log U}{\log q}}(x) &= P\left(\frac{\log U}{\log q} \leq x\right) \\ &= P(\log U \leq \log q^x) \\ &= P(U \leq q^x) \end{aligned}$$

$$\text{For pmf, } P(U = x) = F_U(x) - F_U(x-1)$$

$$\begin{aligned} \text{Similarly, } P\left(\frac{\log U}{\log q} = x\right) &= P(U \leq q^x) - P(U \leq q^{x-1}) \\ &= q^x - q^{x-1} = q^{x-1}(1-q), \text{ for } x > 0. \end{aligned}$$

Therefore,  $P_{\frac{\log U}{\log q}}(x)$  follows a geometric distribution as  $P(\frac{\log U}{\log q} = x) = q^{x-1}(1-q)$  is the probability mass function of a geometric distribution with parameter  $p$  by definition.

Moreover, when we take lower bound of  $\frac{\log U}{\log q}$ , we shift the range of  $\frac{\log U}{\log q}$  to positive integers greater than or equal to  $\lfloor \frac{\log U}{\log q} \rfloor$ . When we add 1 to  $\lfloor \frac{\log U}{\log q} \rfloor$ , we further shift the set of random variable one unit to the positive side. The shift itself makes no difference to the type of distribution, so  $X = 1 + \lfloor \frac{\log U}{\log q} \rfloor$  follows a geometric distribution with the geometric parameter equaling  $q$ .

2. Bayes rule.

Bayes' Rule:

$$P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)}$$

(a)  $n=2$

By Bayes' Rule,

$$P(N=1 | Z=0.05) = \frac{P(Z=0.05 | N=1) P(N=1)}{P(Z=0.05)}$$

For  $P(Z=0.05)$ , it means the probability of having all  $N$  iid uniform RV  $\{X_i\}_{i=1, \dots, N}$  greater than 0.05.

Simply considering  $N=1$ :

$$\begin{aligned} P(Z=0.05 | N=1) &= P(X > 0.05) \\ &= 1 - P(X \leq 0.05) \end{aligned}$$

By cumulative density function of uniform distribution,

$$F(x) = P(X \leq x) = \begin{cases} x & \text{for } 0 \leq x \leq 1 \\ 0 & \text{for } x < 0 \\ 1 & \text{for } x > 1 \end{cases}$$

$$\text{Therefore, } P(Z=0.05 | N=1) = 1 - 0.05 = 0.95$$

Then, for  $P(Z=0.05)$ , we can use law of total probability.

$$\text{When } n=2, \quad P(N=1) = P(N=2) = \frac{1}{2}$$

$$\begin{aligned} P(Z=0.05) &= P(Z=0.05 | N=1) P(N=1) + P(Z=0.05 | N=2) P(N=2) \\ &= 0.95 \cdot \frac{1}{2} + P(Z=0.05 | N=2) \cdot \frac{1}{2} \end{aligned}$$

When there are 2 R.V.s generated from  $\text{unif} \sim (0, 1)$ .

$$P(Z=0.05 | N=2) = P(X_1 > 0.05 \text{ and } X_2 > 0.05).$$

Since  $X_1$  and  $X_2$  are independent and identical R.V.s.

$$\begin{aligned} &= P(X_1 > 0.05) \cdot P(X_2 > 0.05) \\ &= (1 - P(X \leq 0.05))^2 \\ &= 0.95^2 \end{aligned}$$

$$\text{Thus, } P(Z=0.05) = (0.95 + 0.95^2) \frac{1}{2}$$

$$\text{Above all, } P(N=1 | Z=0.05) = \frac{0.95 \cdot \cancel{\frac{1}{2}}}{(0.95 + 0.95^2) \cdot \cancel{\frac{1}{2}}} = 0.5128$$

(b)  $n = 10$ .

Similarly,

By Bayes' Rule,

$$P(N=1 | Z=0.05) = \frac{P(Z=0.05 | N=1) P(N=1)}{P(Z=0.05)}$$

In this case, still,  $P(Z=0.05 | N=1) = 0.95$

$$P(N=1) = P(N=10) = 0.5.$$

$$\begin{aligned} \text{However, } P(Z=0.05) &= P(Z=0.05 | N=1) P(N=1) + P(Z=0.05 | N=10) P(N=10) \\ &= 0.95 \times \frac{1}{2} + P(Z=0.05 | N=10) \times \frac{1}{2} \end{aligned}$$

Considering the second term:

$$P(Z=0.05 | N=10) = P(X_1 > 0.05, X_2 > 0.05, \dots, X_{10} > 0.05)$$

Due to the reason that  $X_i$  from  $i=1$  to  $10$  are all iid,

$$= \prod_{i=1}^{10} P(X_i > 0.05)$$

$$= \prod_{i=1}^{10} (1 - P(X_i \leq 0.05))$$

Still, by cdf of uniform distribution,

$$= \prod_{i=1}^{10} (1 - 0.05)$$

$$= 0.95^{10}$$

$$\text{Therefore, } P(Z=0.05) = (0.95 + 0.95^{10}) \cdot \frac{1}{2}$$

$$\begin{aligned} \text{In summary, } P(N=1 | Z=0.05) &= \frac{0.95 \cdot \cancel{0.5}}{(0.95 + 0.95^{10}) \cdot \cancel{0.5}} \\ &= \frac{1}{1 + 0.95^9} \end{aligned}$$

$$\approx 0.6134$$