

01.

(a).

$$E[S^x] = E[E(S^x | U)] = \int_0^1 (1 + u(s-1))^n du = \frac{1}{n+1} \cdot \frac{1-s^{n+1}}{1-s},$$

which is the prob. generating fⁿ of the uniform distⁿ with $a=0$, $b=n$.

(b) for $|s| < \mu+1$,

$$\begin{aligned} E[S^x] &= E[E(S^x | \Lambda)] = E[e^{\Lambda(s-1)}] = \frac{\mu}{1+\mu-s} \\ &= \frac{\mu}{1-\frac{s}{1+\mu}} \cdot \frac{1}{1+\mu} = \frac{\mu}{1-\frac{s}{1+\mu}}, \end{aligned}$$

which is the prob. generating fⁿ of a Type I Neg. Bin with $k=1$, \Rightarrow A geometric distⁿ.

Q2.

(a) Given $X = \sigma \rho U + \sigma \sqrt{1-\rho^2} V$,

$$Y = \tau U.$$

$$E[X|Y=y] = E[\sigma \rho U | U=y/\tau] = \frac{\sigma \rho y}{\tau}$$

$$\begin{aligned} (b) E[X^2|Y=y] &= E[(\sigma \rho U)^2 + \sigma^2(1-\rho^2)V^2 | U=y/\tau] \\ &= \left(\frac{\sigma \rho y}{\tau}\right)^2 + \sigma^2(1-\rho^2) \end{aligned}$$

$$\Rightarrow \text{Var}(X|Y=y) = \sigma^2(1-\rho^2)$$

(c) Use the conclusion that If X & Y are indep. Normal,
then $E[X|X+Y=z] = az + b$, where
$$\begin{cases} a = \frac{\text{Cov}(X, Z)}{\sigma_Z^2} \\ b = \mu_X - a\mu_Z \end{cases}$$

$$\text{Cov}(X, X+Y) = \text{Var}(X) + \text{Cov}(X, Y) = \sigma^2 + \rho\sigma\tau$$

$$\begin{aligned} \text{Var}(Z) &= \text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y) \\ &= \sigma^2 + \tau^2 + 2\rho\sigma\tau \end{aligned}$$

$$E[X|Z] = \frac{(\sigma^2 + \rho\sigma\tau)z}{\sigma^2 + \tau^2 + 2\rho\sigma\tau}$$

$$(d) 1 - \rho(X, X+Y)^2 = \frac{\tau^2(1-\rho^2)}{\sigma^2 + \tau^2 + 2\rho\sigma\tau}$$

$$\Rightarrow \text{Var}(X|Z) = \frac{\sigma^2\tau^2(1-\rho^2)}{\tau^2 + \sigma^2 + 2\rho\sigma\tau}$$

Q3.

(a) Since V is symmetric, there exists a non-singular matrix M such that $M' = M^{-1}$ and $V = M \Lambda M^{-1}$, where Λ is the diagonal matrix with diagonal entries the eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ of V . Let $\Lambda^{\frac{1}{2}}$ be the diagonal matrix with diag. entries $\lambda_1^{\frac{1}{2}}, \lambda_2^{\frac{1}{2}}, \dots, \lambda_n^{\frac{1}{2}}$; $\Lambda^{\frac{1}{2}}$ is well defined since V is non-negative definite. Writing $W = M \Lambda^{\frac{1}{2}} M^{-1}$, we have that $W = W'$ and also

$$W^2 = (M \Lambda^{\frac{1}{2}} M^{-1})(M \Lambda^{\frac{1}{2}} M^{-1}) = M \Lambda M^{-1} = V.$$

Students obtain full marks for the above. for Q3.

Clearly W is non-singular iff $\Lambda^{\frac{1}{2}}$ is non-singular.

This happens iff $\lambda_i > 0$ for all i , which is to say that

V is positive definite.

Q4.

$$\begin{aligned} a) E[x^{x+y}] &= E\{E[x^{x+y} | y]\} = E[x^y e^{y(x-1)}] \\ &= E[(xe^{x-1})^y] = \exp\{\mu(xe^{x-1}-1)\} \end{aligned}$$

(b) To obtain the MGF for x firstly.

$$\begin{aligned} M_x(t) &= \sum_{x=1}^{\infty} e^{xt} \frac{(1-p)^x}{x \ln(1/p)} \\ &= \sum_{x=1}^{\infty} \frac{[e^t(1-p)]^x}{x \ln(1/p)} = \sum_{x=1}^{\infty} \frac{[1 - (1 - e^t(1-p))]^x \ln(1 - e^t(1-p))}{x \ln(1/p) \cdot \ln(1 - e^t(1-p))} \\ &= \frac{\ln(1 - e^t(1-p))}{\ln(1/p)} \end{aligned}$$

$$\begin{aligned} M_s(t) &= \exp[\mu(M_x(t) - 1)] \\ &= \exp\left[\mu \left\{ \frac{\ln(1 - e^t(1-p)) - \ln(1/p)}{\ln(1/p)} \right\}\right] \\ &= \exp\left[\mu \cdot \frac{\ln \frac{p}{1 - e^t(1-p)}}{\ln \frac{1}{p}}\right] = \left\{ e^{\ln \frac{p}{1 - e^t(1-p)}} \right\}^{\mu \frac{1}{\ln \frac{1}{p}}} \end{aligned}$$

$$= \left[\left(\frac{p}{1 - e^{t(1-p)}} \right)^u \right]_{\ln \frac{1}{p}}^{\frac{1}{p}} = \left[\frac{1 - (1-p)^{\frac{1}{p}}}{1 - e^{t(1-p)}} \right]_{\ln \frac{1}{p}}^{\frac{1}{p}},$$

which is the MGF of a NB.

$$\begin{aligned} (c) E\left[\frac{1}{1+X}\right] &= E\left[\int_0^1 t^X dt\right] = \int_0^1 E[t^X] dt \\ &= \int_0^1 (q + pt)^n dt = \frac{1 - q^{n+1}}{p(n+1)}. \end{aligned}$$