

Problem 1

(a) Let  $X \in \text{Exp}(1)$  and  $Y \in \text{Exp}(1)$  be independent random variables

Do  $\frac{X+Y}{2}$  and  $\max(X, Y)$  have the same distribution?

Fix  $u \geq 0$ .

$$\begin{aligned} \mathbb{P}\left(\frac{X+Y}{2} \leq u\right) &= \mathbb{P}(X+Y \leq 2u) = \int_0^{2u} f_{X+Y}(v) dv = \int_0^{2u} \int_0^{2u-v} f_X(v-y) f_Y(y) dy dv \\ &= \int_0^{2u} \int_0^{2u-v} e^{-(v-y)} \mathbb{1}_{\{v-y \geq 0\}} e^{-y} \mathbb{1}_{\{y \geq 0\}} dy dv = \int_0^{2u} \int_0^v e^{-v} dy dv = \int_0^{2u} e^{-v} \cdot v dv = -e^{-v} \cdot v \Big|_0^{2u} + \int_0^{2u} e^{-v} dv \\ &= -e^{-2u} \cdot 2u + (-e^{-v}) \Big|_0^{2u} = -e^{-2u} \cdot 2u + 1 - e^{-2u} = 1 - e^{-2u}(1+2u) \end{aligned}$$

Now  $\mathbb{P}(\max(X, Y) \leq u) = \mathbb{P}(\{X \leq u\} \cap \{Y \leq u\}) \stackrel{\text{ind}}{=} \mathbb{P}(X \leq u) \mathbb{P}(Y \leq u) = \left(\int_0^u e^{-x} dx\right)^2 = (-e^{-x}) \Big|_0^u = (1 - e^{-u})^2$

$\Rightarrow \mathbb{P}\left(\frac{X+Y}{2} \leq u\right) \neq \mathbb{P}(\max(X, Y) \leq u)$ , so they are not identically distributed.

(b) Let  $X_1, X_2, \dots, X_n, \dots$  be iid  $\mathcal{N}(\mu, \sigma^2)$ -random variables with  $\mu \neq 0$ .

To determine the asymptotic distribution of  $\left(\left(\frac{1}{n} \sum_{i=1}^n X_i\right)^2 - \mu^2\right) \sqrt{n}$ .

We want to use the law of large numbers and the central limit theorem:

$$\begin{aligned} \sqrt{n} \left( \left(\frac{1}{n} \sum_{i=1}^n X_i\right)^2 - \mu^2 \right) &= \sqrt{n} \underbrace{\left(\frac{1}{n} \sum_{i=1}^n X_i - \mu\right)}_{\xrightarrow{\mathbb{P}} \mathcal{N}(0, \sigma^2) \text{ as } n \rightarrow \infty} \underbrace{\left(\frac{1}{n} \sum_{i=1}^n X_i + \mu\right)}_{\xrightarrow{\mathbb{P}} 2\mu \text{ as } n \rightarrow \infty \text{ (by the law of large numbers)}} \\ &\quad \text{(by the central limit theorem)} \end{aligned}$$

By Slutsky,  $\sqrt{n} \left( \left(\frac{1}{n} \sum_{i=1}^n X_i\right)^2 - \mu^2 \right) \xrightarrow{\mathbb{P}} 2\mu \cdot \mathcal{N}(0, \sigma^2) = \mathcal{N}(0, 4\mu^2 \sigma^2)$  as  $n \rightarrow \infty$

$\Rightarrow \sqrt{n} \left( \left(\frac{1}{n} \sum_{i=1}^n X_i\right)^2 - \mu^2 \right) \xrightarrow{d} \mathcal{N}(0, 4\mu^2 \sigma^2)$  as  $n \rightarrow \infty$ .

Problem 2: Let  $(X_n)_{n \geq 1}$ ;  $\mathbb{P}(X_n = 1) = \mathbb{P}(X_n = -1) = \frac{1}{2}$ .

Let  $N \in \mathcal{P}_0(\lambda)$  and independent of  $(X_n)_{n \geq 1}$ . Let  $S_N = X_1^2 + X_2^2 + \dots + X_N^2$ ,  $S_0 = 0$ .

To show:  $\frac{S_N}{\sqrt{\lambda}} \xrightarrow{d} \mathcal{N}(0, 1)$  as  $\lambda \rightarrow \infty$ .

By the course,  $\left\{ \frac{S_N}{\sqrt{\lambda}} \xrightarrow{d} \mathcal{N}(0, 1) \text{ as } \lambda \rightarrow \infty \right\} \Leftrightarrow \left\{ \mathbb{E} \left[ e^{it \frac{S_N}{\sqrt{\lambda}}} \right] \rightarrow e^{-t^2/2} \text{ as } \lambda \rightarrow \infty \right\}$   
 $\forall t \in \mathbb{R}$ .

$$\mathbb{E} \left[ e^{it \frac{S_N}{\sqrt{\lambda}}} \right] = \mathbb{E} \left[ \mathbb{E} \left[ e^{it \frac{S_N}{\sqrt{\lambda}}} \mid N \right] \right] \stackrel{\text{ind. of } (X_j)}{=} \mathbb{P}(N=0) + \sum_{k=1}^{\infty} \mathbb{E} \left[ e^{it \frac{S_N}{\sqrt{\lambda}}} \right] \mathbb{P}(N=k) = \mathbb{P}(N=0) + \sum_{k=1}^{\infty} \mathbb{E} \left[ \prod_{j=1}^k e^{it X_j^2 / \sqrt{\lambda}} \right] \mathbb{P}(N=k)$$

$$\begin{aligned} &\stackrel{(X_j) \text{ iid}}{=} \mathbb{P}(N=0) + \sum_{k=1}^{\infty} \left( \mathbb{E} \left[ e^{it X_1^2 / \sqrt{\lambda}} \right] \right)^k \mathbb{P}(N=k) = \mathbb{P}(N=0) + \sum_{k=1}^{\infty} \left( \mathbb{E} \left[ e^{it X_1^2 / \sqrt{\lambda}} \right] \right)^k \mathbb{P}(N=k) \quad (*) \\ &\quad X_1^2 = X_2^2 \text{ since } 1^2 = 1 \text{ and } (-1)^2 = 1. \end{aligned}$$

Now,  $E[e^{itX_1/\sqrt{n}}] = e^{it/\sqrt{n}} \cdot \frac{1}{2} + e^{-it/\sqrt{n}} \cdot \frac{1}{2} = \cos(t/\sqrt{n})$

$\Rightarrow (*) = e^{-n} + \sum_{k=2}^{\infty} (\cos(t/\sqrt{n}))^k e^{-n} \frac{n^k}{k!} = e^{-n} \left( 1 + \sum_{k=2}^{\infty} \frac{(\cos(t/\sqrt{n}) \cdot n)^k}{k!} \right) = e^{-n} \cdot e^{\cos(t/\sqrt{n}) \cdot n} = e^{n(\cos(t/\sqrt{n}) - 1)}$

Now we want to determine  $\lim_{n \rightarrow \infty} n(\cos(t/\sqrt{n}) - 1)$ .  $n(\cos(t/\sqrt{n}) - 1) = n \left( \sum_{k=2}^{\infty} \frac{(t/\sqrt{n})^k}{(2k)!} (-1)^k - 1 \right)$

$= n \sum_{k=2}^{\infty} \frac{(t/\sqrt{n})^{2k}}{(2k)!} \cdot (-1)^k = \frac{-t^2}{2} + \sum_{k=2}^{\infty} n \frac{(t/\sqrt{n})^{2k}}{(2k)!} \cdot (-1)^k$   
 $\Rightarrow \lim_{n \rightarrow \infty} E[e^{itX_1/\sqrt{n}}] = e^{-t^2/2}$ , the characteristic function of a  $N(0,1)$ -distributed random variable.

Problem 3 Let  $X = \begin{pmatrix} X_1 \\ X_2 \end{pmatrix} \in N(\mu, \Lambda)$ , where  $\mu = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$  and  $\Lambda = \begin{pmatrix} 3 & 1 \\ 1 & 2 \end{pmatrix}$

(1) compute  $P\{X_1 \geq 2 \mid X_2 + 3X_1 = 3\}$

Let's first compute  $f_{X_1 | 3X_1 + X_2 = 3}(x_1) = \frac{f(x_1, 3x_1 + x_2)(x_1, 3)}{f_{3X_1 + X_2}(3)}$

Lemma (converse):  $X \in N(\mu, \Lambda) \Leftrightarrow a_1 X_1 + a_2 X_2$  is normally distributed,  $\forall a_1, a_2 \in \mathbb{R}$

$\Rightarrow 3X_1 + X_2$  is normally distributed and  $\begin{pmatrix} X_1 \\ 3X_1 + X_2 \end{pmatrix}$  is normally distributed.

Hence we just have to compute the covariance matrix and the mean vector of  $\begin{pmatrix} X_1 \\ 3X_1 + X_2 \end{pmatrix}$  to determine the law.

Reminder: Let  $A$  be a  $2 \times 2$ -matrix.  $E[A \begin{pmatrix} X_1 \\ X_2 \end{pmatrix}] = A E[\begin{pmatrix} X_1 \\ X_2 \end{pmatrix}]$  by linearity

$Cov(A \begin{pmatrix} X_1 \\ X_2 \end{pmatrix}) = E[(A \begin{pmatrix} X_1 \\ X_2 \end{pmatrix} - E[A \begin{pmatrix} X_1 \\ X_2 \end{pmatrix}]) (A \begin{pmatrix} X_1 \\ X_2 \end{pmatrix} - E[A \begin{pmatrix} X_1 \\ X_2 \end{pmatrix}])^T]$

$= E[A ((\begin{pmatrix} X_1 \\ X_2 \end{pmatrix} - E[\begin{pmatrix} X_1 \\ X_2 \end{pmatrix}])) ((\begin{pmatrix} X_1 \\ X_2 \end{pmatrix} - E[\begin{pmatrix} X_1 \\ X_2 \end{pmatrix}]))^T A^T] = A Cov(\begin{pmatrix} X_1 \\ X_2 \end{pmatrix}) A^T$

Now  $\begin{pmatrix} X_1 \\ 3X_1 + X_2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} X_1 \\ X_2 \end{pmatrix} \Rightarrow E[\begin{pmatrix} X_1 \\ 3X_1 + X_2 \end{pmatrix}] = \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} E[\begin{pmatrix} X_1 \\ X_2 \end{pmatrix}] = \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 \\ 4 \end{pmatrix}$

and  $Cov(\begin{pmatrix} X_1 \\ 3X_1 + X_2 \end{pmatrix}) = Cov(\begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} X_1 \\ X_2 \end{pmatrix}) = \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} Cov(\begin{pmatrix} X_1 \\ X_2 \end{pmatrix}) \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix}^T = \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} Cov(\begin{pmatrix} X_1 \\ X_2 \end{pmatrix}) \begin{pmatrix} 1 & 3 \\ 0 & 1 \end{pmatrix}$

$= \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} 3 & 1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} 1 & 3 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} 2 & 10 \\ 1 & 5 \end{pmatrix} = \begin{pmatrix} 3 & 10 \\ 10 & 35 \end{pmatrix}$

$\Rightarrow \begin{pmatrix} X_1 \\ 3X_1 + X_2 \end{pmatrix} \in N(\begin{pmatrix} 1 \\ 4 \end{pmatrix}, \begin{pmatrix} 3 & 10 \\ 10 & 35 \end{pmatrix})$  and  $3X_1 + X_2 \in N(4, 35)$

$\Rightarrow f(x_1, 3x_1 + x_2)(x_1, 3) = \frac{1}{2\pi \sqrt{5}} e^{-\frac{1}{2} \left( \begin{pmatrix} x_1 \\ 3 \end{pmatrix} - \begin{pmatrix} 1 \\ 4 \end{pmatrix} \right)^T \frac{1}{5} \begin{pmatrix} 35 & -10 \\ -10 & 3 \end{pmatrix} \left( \begin{pmatrix} x_1 \\ 3 \end{pmatrix} - \begin{pmatrix} 1 \\ 4 \end{pmatrix} \right)}$

$\det \begin{pmatrix} 3 & 10 \\ 10 & 35 \end{pmatrix} = 5$

and  $\left( \begin{pmatrix} x_1 \\ 3 \end{pmatrix} - \begin{pmatrix} 1 \\ 4 \end{pmatrix} \right)^T \frac{1}{5} \begin{pmatrix} 35 & -10 \\ -10 & 3 \end{pmatrix} \left( \begin{pmatrix} x_1 \\ 3 \end{pmatrix} - \begin{pmatrix} 1 \\ 4 \end{pmatrix} \right) =$

$\begin{pmatrix} 3 & 10 \\ 10 & 35 \end{pmatrix}^{-1} = \frac{1}{5} \begin{pmatrix} 35 & -10 \\ -10 & 3 \end{pmatrix}$

$\begin{pmatrix} x_1 - 1 & -1 \end{pmatrix} \frac{1}{5} \begin{pmatrix} 35 & -10 \\ -10 & 3 \end{pmatrix} \begin{pmatrix} x_1 - 1 \\ -1 \end{pmatrix} = 7(x_1 - 1)^2 + 4(x_1 - 1) + \frac{3}{5}$   
 $= 7x_1^2 - 10x_1 + 1 + \frac{3}{5} = 7x_1^2 - 10x_1 + \frac{18}{5}$

$$\Rightarrow f_{(X_1, 3X_1+X_2)}(x_1, z) = \frac{1}{2\pi\sqrt{5}} e^{-\frac{1}{2}(7x_1^2 - 10x_1 + \frac{16}{5})}$$

$$f_{3X_1+X_2}(z) = \frac{1}{\sqrt{2\pi} \cdot 1.85} e^{-\frac{1}{2}(\frac{(z-4)^2}{35})} = \frac{1}{\sqrt{2\pi} \cdot 35} e^{-\frac{1}{2}(\frac{z}{35})}$$

Now  $7x_1^2 - 10x_1 + \frac{16}{5} - \frac{1}{35} = 7x_1^2 - 10x_1 + \frac{125}{35} = 7x_1^2 - 10x_1 + \frac{25}{7} = 7(x_1 - \frac{5}{7})^2$

$$\Rightarrow f_{X_1|3X_1+X_2=3}(x_1) = \frac{1}{\sqrt{2\pi} \cdot \frac{1}{\sqrt{7}}} e^{-\frac{1}{2}(\frac{(x_1 - 5/7)^2}{1/7})} \Rightarrow X_1|3X_1+X_2=3 \in \mathcal{N}(\frac{5}{7}, \frac{1}{7})$$

Now we want to express  $P(X_1 \geq 2 | 3X_1 + X_2 = 3)$  as a function of  $\Phi(t) := P(X \leq t)$ , where  $X \in \mathcal{N}(0, 1)$ .

Let  $Y \in \mathcal{N}(\frac{5}{7}, \frac{1}{7})$ .  $P(X_1 \geq 2 | 3X_1 + X_2 = 3) = P(Y \geq 2) = P(\frac{Y - \frac{5}{7}}{\frac{1}{\sqrt{7}}} \geq \frac{2 - 5/7}{\frac{1}{\sqrt{7}}}) = P(X \geq 9/\sqrt{7}) = 1 - \Phi(\frac{9}{\sqrt{7}})$

(2) Find  $E[X_1^2 X_2^8 | X_2 = 2]$

As before, we first need to compute  $\frac{f_{(X_1, X_2)}(x_1, 2)}{f_{X_2}(2)}$  ( $(X_1, X_2) \in \mathcal{N}(\begin{pmatrix} 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 3 & 1 \\ 1 & 2 \end{pmatrix})$  and  $X_2 \in \mathcal{N}(2, 2)$ )

$$f_{X_1|X_2=2}(x_1) \stackrel{\text{def}}{=} \frac{1}{2\pi\sqrt{5}} e^{-\frac{1}{2} \left( \left( \begin{pmatrix} x_1 \\ 2 \end{pmatrix} - \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right)^T \frac{1}{5} \begin{pmatrix} 2 & -1 \\ -1 & 3 \end{pmatrix} \left( \begin{pmatrix} x_1 \\ 2 \end{pmatrix} - \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right) \right)} = \frac{\sqrt{2}}{\sqrt{2\pi} \sqrt{5}} e^{-\frac{1}{2} \left( \frac{1}{5} (2x_1^2 - 6x_1 + 7) \right) - \frac{1}{2}}$$

$$\frac{1}{\sqrt{2\pi} \sqrt{2}} e^{-\frac{1}{2} \left( \frac{(x_1 - 2)^2}{2} \right)}$$

$$\left( \begin{pmatrix} x_1 \\ 2 \end{pmatrix} - \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right)^T \frac{1}{5} \begin{pmatrix} 2 & -1 \\ -1 & 3 \end{pmatrix} \left( \begin{pmatrix} x_1 \\ 2 \end{pmatrix} - \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right) =$$

$$\frac{1}{5} (2(x_1 - 1)^2 - 2(x_1 - 1)(x_1 - 1) + 3(2 - 2)^2) =$$

$$\frac{1}{5} (2x_1^2 - 4x_1 + 2 - 2x_1 + 2 + 3) = \frac{1}{5} (2x_1^2 - 6x_1 + 7)$$

$$= \frac{\sqrt{2}}{\sqrt{2\pi} \sqrt{5}} e^{-\frac{1}{2} \left( \frac{1}{5} (2x_1^2 - 6x_1 + 7 - \frac{5}{2}) \right)} = \frac{\sqrt{2}}{\sqrt{2\pi} \sqrt{5}} e^{-\frac{1}{2} \left( \frac{2}{5} (x_1^2 - 3x_1 + \frac{9}{4}) \right)} = \frac{\sqrt{2}}{\sqrt{2\pi} \sqrt{5}} e^{-\frac{1}{2} \left( \frac{2}{5} (x_1 - \frac{3}{2})^2 \right)}$$

Now  $E[X_1^2 X_2^8 | X_2 = 2] = E[X_1^2 \cdot 2^8 | X_2 = 2] = 2^8 E[X_1^2 | X_2 = 2] = 2^8 (\text{Var}(X_1 | X_2 = 2) + E[X_1 | X_2 = 2]^2)$

$$= 2^8 \left( \frac{5}{2} + \frac{9}{4} \right) = 2^6 \cdot 19$$

We know these two quantities by the computation of  $f_{X_1|X_2=2}(x_1)$

Remark:  $U|V=k \sim \mathcal{N}(\mu_U - \frac{Cov}{Cv}, \sigma_U^2 - \frac{Cov^2}{Cv})$

Problem 3 (Exam 2017)

Let  $X_1, X_2, \dots$  be  $U(0,1)$ -distributed i.i.d. random variables and  $N \in \mathcal{P}_0(\lambda)$  be independent of  $X_1, X_2, \dots$

Set  $Y_N = \max\{X_1, X_2, \dots, X_N\}$  and  $Y_0 = 0$ .

(a.) Fix  $u \in (0,1)$ . 
$$\mathbb{P}(Y_N \leq u) = \sum_{k \geq 0} \mathbb{P}(Y_N \leq u | N=k) \mathbb{P}(N=k) = \sum_{k \geq 0} \mathbb{P}(Y_k \leq u | N=k) \mathbb{P}(N=k)$$

At ind. 
$$= \sum_{k \geq 0} \mathbb{P}(Y_k \leq u) \mathbb{P}(N=k) = \underbrace{\mathbb{P}(0 \leq u)}_{=1 \text{ since } u \in (0,1)} \mathbb{P}(N=0) + \sum_{k \geq 1} \mathbb{P}(Y_k \leq u) \mathbb{P}(N=k) = \mathbb{P}(N=0) + \sum_{k \geq 1} \mathbb{P}(\max\{X_1, \dots, X_k\} \leq u) \mathbb{P}(N=k)$$

$$= \mathbb{P}(N=0) + \sum_{k \geq 1} \mathbb{P}(X_i \leq u \forall i \leq k) \mathbb{P}(N=k) = \mathbb{P}(N=0) + \sum_{k \geq 1} \mathbb{P}(X_1 \leq u)^k \mathbb{P}(N=k) = e^{-\lambda} + \sum_{k \geq 1} u^k \frac{\lambda^k e^{-\lambda}}{k!}$$

$$= e^{-\lambda} \left( 1 + \sum_{k \geq 1} \frac{(u\lambda)^k}{k!} \right) = e^{-\lambda} \cdot e^{u\lambda} = e^{-\lambda(1-u)}$$

Now for the characteristic function: we want to find  $f_{Y_N}(u)$  first

$$f_{Y_N}(u) = \frac{d}{du} \mathbb{P}(Y_N \leq u) = \frac{d}{du} e^{-\lambda(1-u)} = e^{-\lambda(1-u)} \cdot \lambda$$

Now notice that  $\mathbb{P}(Y_N=0) = \mathbb{P}(N=0) = e^{-\lambda} > 0$ .

Hence 
$$\mathbb{E}[e^{itY_N}] = \mathbb{E}[e^{itY_N} \mathbb{1}_{\{N=0\}}] + \mathbb{E}[e^{itY_N} \mathbb{1}_{\{N>0\}}] = \mathbb{P}(N=0) + \int_0^1 e^{itu} \lambda e^{-\lambda(1-u)} du =$$

$$e^{-\lambda} + \int_0^1 \lambda e^{itu - \lambda(1-u)} du = e^{-\lambda} \left( 1 + \int_0^1 \lambda e^{u(it+\lambda)} du \right) = e^{-\lambda} \left( 1 + \frac{\lambda}{it+\lambda} e^{u(it+\lambda)} \Big|_0^1 \right)$$

$$= e^{-\lambda} \left( 1 + \frac{\lambda}{it+\lambda} e^{it+\lambda} - \frac{\lambda}{it+\lambda} \right)$$

(b.) 
$$\mathbb{E}[Y_N] = \mathbb{E}[Y_N \mathbb{1}_{\{N=0\}}] + \mathbb{E}[Y_N \mathbb{1}_{\{N>0\}}] = \mathbb{P}(N=0) + \int_0^1 \mathbb{P}(Y_N > t) dt = \mathbb{P}(N=0) + \int_0^1 (1 - \mathbb{P}(Y_N \leq t)) dt$$

$$= e^{-\lambda} + \int_0^1 (1 - e^{-\lambda(1+t)}) dt \xrightarrow{\lambda \rightarrow \infty} 1 + 0 = 1$$

(c.) Show that  $\lambda(1-Y_N)$  converges in distribution as  $\lambda \rightarrow +\infty$  and determine the limiting distribution.

$$\mathbb{P}(\lambda(1-Y_N) \leq u) = \mathbb{P}(\lambda - \lambda Y_N \leq u) = \mathbb{P}(-\lambda Y_N \leq u - \lambda) = \mathbb{P}(Y_N \geq \frac{\lambda - u}{\lambda}) = 1 - \mathbb{P}(Y_N \leq \frac{\lambda - u}{\lambda})$$

$$= 1 - e^{-\lambda(1 - \frac{\lambda - u}{\lambda})} = 1 - e^{-\lambda(\frac{u}{\lambda})} = 1 - e^{-u} = \mathbb{P}(X \leq u), \text{ where } X \in \text{Exp}(1)$$

Hence it finally converges to  $\text{Exp}(1)$  as  $\lambda \rightarrow +\infty$ .

AED

Problem 4

Let  $\varphi(t,s,u) = \exp\{2is - s^2 - 2t^2 - 4u^2 - 2st + 2su\}$  be the characteristic function of  $\begin{pmatrix} X \\ Y \\ Z \end{pmatrix}$ .

Determine the distribution of  $\begin{pmatrix} X \\ Y \\ Z \end{pmatrix}$ .

Let  $\vec{V}$  be a Gaussian vector. Then  $\varphi_{\vec{V}}(t,s,u) = \exp\{i(tsu) \cdot \mu - \frac{1}{2}(tsu) C \begin{pmatrix} t \\ s \\ u \end{pmatrix}\}$ .

We are going to find  $\mu$  and  $C$  so that  $\varphi(t,s,u) = \varphi_{\vec{V}}(t,s,u)$ .

$$\varphi_{\vec{V}}(t,s,u) = \exp\{it\mu_1 + is\mu_2 + iu\mu_3 - \frac{1}{2}(c_{11}t^2 + c_{22}s^2 + c_{33}u^2 + 2c_{12}ts + 2c_{23}su + 2c_{13}tu)\}$$

$$\Rightarrow \mu_1 = 0, \mu_2 = 2, \mu_3 = 0, c_{11} = 4, c_{22} = -2, c_{33} = 8, c_{12} = 2, c_{13} = 0, c_{23} = -2.$$

$$\Rightarrow \begin{pmatrix} X \\ Y \\ Z \end{pmatrix} \in \mathcal{N}\left(\begin{pmatrix} 0 \\ 2 \\ 0 \end{pmatrix}, \begin{pmatrix} 4 & 2 & 0 \\ 2 & -2 & -2 \\ 0 & -2 & 8 \end{pmatrix}\right)$$

MORELLOM Philippe (X2940) (Additional Exercise)

[T.X] 3.8.3.11 Let  $X \in \text{Exp}(1)$ ,  $Y \in \text{Exp}(1)$  be independent. To show:  $\frac{X}{X+Y} \in U(0,1)$ .

Fix  $u \geq 0$ .

$$\mathbb{P}\left(\frac{X}{X+Y} \leq u\right) = \mathbb{P}(X \leq u(X+Y)) = \mathbb{P}(X(1-u) \leq uY) = \mathbb{P}\left(X \leq \frac{u}{1-u} Y\right) \quad (\text{Notice that } u < 1, \text{ otherwise } \frac{u}{1-u} \leq 0 \text{ and } X \text{ is negative! which has } \mathbb{P} = 0)$$

$X+Y \geq 0$  since they are  $\text{Exp}(1)$

$$= \int_0^{+\infty} \int_0^{+\infty} \mathbb{1}_{\left\{X \leq \frac{u}{1-u} Y\right\}} f_X(x) f_Y(y) dx dy = \int_0^{+\infty} \int_0^{\frac{u}{1-u} y} f_X(x) f_Y(y) dx dy = \int_0^{+\infty} \int_0^{\frac{u}{1-u} y} e^{-x} e^{-y} dx dy$$

$$= \int_0^{+\infty} e^{-y} \left\{ \int_0^{\frac{u}{1-u} y} e^{-x} dx \right\} dy = \int_0^{+\infty} e^{-y} \left\{ -e^{-x} \Big|_0^{\frac{u}{1-u} y} \right\} dy = \int_0^{+\infty} e^{-y} \left( 1 - e^{-\frac{u}{1-u} y} \right) dy$$

$$= \int_0^{+\infty} e^{-y} dy - \int_0^{+\infty} e^{-y} \left( 1 + \frac{u}{1-u} \right) dy = \int_0^{+\infty} e^{-y} dy - \int_0^{+\infty} e^{-y} \left( \frac{1}{1-u} \right) dy$$

$$= -e^{-y} \Big|_0^{+\infty} + (1-u) e^{-y} \left( \frac{1}{1-u} \right) \Big|_0^{+\infty} = 1 - (1-u) = u \quad \forall u \in (0,1).$$

$$\Rightarrow \frac{X}{X+Y} \in U(0,1) !$$

QED