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• For a random vector Y , the characteristic function (CF), denoted by $\phi_Y(t)$, is defined as

$$E(e^{it^T Y}) \quad \forall t \in \mathbb{R}^n \text{ where } Y \text{ is}$$

a n -dimensional random vector.

It is known that CF of a random vector always exists and uniquely characterizes the distribution.

• Suppose $Y = \begin{pmatrix} Y_1 \\ Y_2 \end{pmatrix}$ is a n -dimensional random vector. Then Y_1 and Y_2 are independent iff

$$\phi_Y(t) = \phi_{Y_1}(t_1) \phi_{Y_2}(t_2) \quad \text{for all } t = (t_1, t_2) \in \mathbb{R}^n$$

• For a random vector Y with

$$Y \sim \text{MVN}(\mu, \Sigma). \text{ Then}$$

$$\text{the } \phi_Y(t) = \underline{e^{(it^T \mu - \frac{1}{2} t^T \Sigma t)}}.$$

(R-1) Suppose $Y \sim \text{MVN}(\mu, \Sigma)$ then for any matrix B .

$$BY \sim \text{MVN}(B\mu, B\Sigma B^T)$$

Proof $\phi_{BY}(t) = E(e^{it^T BY})$

X and Y are rvs.

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

For n rvs

$$P(X_1 \leq x_1, X_2 \leq x_2, \dots, X_n \leq x_n)$$

$$= P(X_1 \leq x_1)$$

$$P(X_n \leq x_n)$$

$$= E \left(e^{i (B^T t)^T Y} \right)$$

$$= E \left(e^{i s^T Y} \right) (= \phi_Y(s)) \quad s = B^T t$$

$$= e^{(i s^T \mu - \frac{1}{2} s^T \Sigma s)}$$

$$= e^{(i t^T B \mu - \frac{1}{2} t^T B \Sigma B^T t)}$$

$$= e^{(i t^T \mu^* - \frac{1}{2} t^T \Sigma^* t)}$$

where $\mu^* = B \mu$

$$BY \sim \text{MVN}(B\mu, B\Sigma B^T) \quad \text{and } \Sigma^* = B\Sigma B^T$$

(R-2)

$$Y \sim \text{MVN}(\mu, \Sigma)$$

$$Y = \begin{pmatrix} Y_1 \\ Y_2 \end{pmatrix}$$

$$\mu = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix} \quad \Sigma = \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}$$

$$\text{Then } Y_1 \sim \text{MVN}(\mu_1, \Sigma_{11}) \quad \checkmark$$

$$Y_2 \sim \text{MVN}(\mu_2, \Sigma_{22})$$

$$B = [I_n \mid 0]$$

$$BY = [I_n \mid 0] \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix} = Y_1$$

$$B\mu = \mu_1 \quad B\Sigma B^T = \Sigma_{11}$$

So, from R-1, we have $Y_1 \sim \text{MVN}(\mu_1, \Sigma_{11})$

$$B = [0 \mid I_{n-n}]$$

$$BY = Y_2, \quad B\mu = \mu_2, \quad \text{and}$$

$$\Sigma = \Sigma^T = \Sigma$$

$$\Sigma_{11} \Sigma_{11}^T = \Sigma_{22}$$

$$Y_2 \sim \text{MVN}(M_2, \Sigma_{22})$$

(R-3) In (R-2) setup, Y_1 and Y_2 are independent
iff $\Sigma_{12} = 0$.

Proof: Suppose Y_1 and Y_2 are independent then
 $\text{cov}(Y_1, Y_2) = 0 \Rightarrow \Sigma_{12} = 0$

Now if $\Sigma_{12} = 0$ then $\Sigma_{21} = 0$

$$\begin{aligned} t^T \Sigma t &= \begin{bmatrix} t_1^T & t_2^T \end{bmatrix} \begin{bmatrix} \Sigma_{11} & 0 \\ 0 & \Sigma_{22} \end{bmatrix} \begin{bmatrix} t_1 \\ t_2 \end{bmatrix} \\ &= t_1^T \Sigma_{11} t_1 + t_2^T \Sigma_{22} t_2 \end{aligned}$$

$$\begin{aligned} \Phi_Y(t) &= e^{(it^T M - \frac{1}{2} t^T \Sigma t)} \\ &= e^{(it_1^T M_1 + it_2^T M_2 - \frac{1}{2} (t_1^T \Sigma_{11} t_1 - \frac{1}{2} (t_2^T \Sigma_{22} t_2))} \\ &= e^{(it_1^T M_1 - \frac{1}{2} (t_1^T \Sigma_{11} t_1))} e^{(it_2^T M_2 - \frac{1}{2} t_2^T \Sigma_{22} t_2)} \\ &= \Phi_{Y_1}(t_1) \Phi_{Y_2}(t_2) \end{aligned}$$

So, Y_1 and Y_2 are independent.

(R-4) $Y \sim \text{MVN}(M, \Sigma)$, AY and BY are
independent iff $A \Sigma B^T = 0$

Proof Suppose AY and BY are independent
then $\text{cov}(AY, BY) = 0$

$$\Rightarrow A \Sigma B^T = 0$$

$$\begin{bmatrix} A \\ B \end{bmatrix} Y \sim \text{MVN} \left(\begin{pmatrix} A\mu \\ B\mu \end{pmatrix}, \begin{pmatrix} A\Sigma A^T & A\Sigma B^T \\ B\Sigma A^T & B\Sigma B^T \end{pmatrix} \right)$$

$$\begin{bmatrix} AY \\ BY \end{bmatrix}$$

Now from (R-3), we can say that

AY and BY are independent if $A\Sigma B^T = 0$.

Fisher's Cochran Theorem

$$A = P^T \begin{bmatrix} I_n & 0 \\ 0 & 0 \end{bmatrix} P$$

$$Y^T A Y = Y^T P^T \begin{bmatrix} I_n & 0 \\ 0 & 0 \end{bmatrix} P Y$$

$$P = \begin{bmatrix} P_1 \\ P_2 \end{bmatrix}$$

$$Y^T A Y = P_1 Y$$

$$\left\{ \begin{array}{l} Y_1, \dots, Y_n \\ \sim N(0, 1) \\ Y^T Y = \sum Y_i^2 \sim \chi_n^2 \\ A \text{ is symmetric and idempotent} \\ Y^T A Y \sim \chi_n^2 \\ \text{where } \pi = P(A) \end{array} \right.$$

$$A = P^T \begin{bmatrix} I_n & 0 \\ 0 & 0 \end{bmatrix} P$$

Theorem Suppose $Y \sim \text{MVN}(0, I_n)$. Further, suppose

$\{A_i\}_{i=1}^k$ are symmetric and idempotent matrices.

with $P(A_i) = \eta_i$. Then if $\sum_{i=1}^k \eta_i = n$, the

following statements are equivalent.

• $A_i A_j = 0 \quad \forall i \neq j$

• $Y^T A_i Y \sim \chi_{\eta_i}^2$ and $Y^T A_i Y$ and $Y^T A_j Y$

are independent.

$$Y^T Y =$$