

### 1.3 Distributions and their characteristics

random  $k$ -vector  $(X_1, \dots, X_k)$  w/ joint pdf  $f$  w.r.t.  $u_1 \times \dots \times u_k$  on  $\mathcal{B}^k$

$\rightarrow$  marginal pdf w.r.t.  $u_i$ :  $f_i(x) = \int_{\mathcal{B}^{k-1}} f(x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_k) du_1 \dots du_{i-1} du_{i+1} \dots du_k$

$X_1, \dots, X_k$  are independent iff  $f(x_1, \dots, x_k) = f_1(x_1) \dots f_k(x_k)$

Lemma:  $X_1, \dots, X_k$  independent  $\rightarrow g(X_1, \dots, X_n), h(X_{n+1}, \dots, X_k)$  independent ( $g, h$  Borel)

Ex: multivariate normal pdf:  $f(x) = (2\pi)^{-k/2} (\det \Sigma)^{-1/2} \exp\left(-\frac{1}{2} (x-\mu)^T \Sigma (x-\mu)\right), x \in \mathbb{R}^k$   
 $\mu \in \mathbb{R}^k, \Sigma$  positive definite  $k \times k$  matrix (if nonnegative definite but singular:  $X$  is  $N_k(\mu, \Sigma)$  iff  $c^T X$  is  $N(c^T \mu, c^T \Sigma c) \forall c \in \mathbb{R}^k$ )  
independent iff  $\Sigma$  is diagonal!

Change of variables:  $X$  random  $k$ -vector w/ Lebesgue pdf  $f_X$ ;  $Y = g(X)$  w/  $g$  Borel from  $(\mathbb{R}^k, \mathcal{B}^k)$  to  $(\mathbb{R}^k, \mathcal{B}^k)$ ;  $A_1, \dots, A_m$  disjoint sets in  $\mathcal{B}^k$  (s.t.  $m(\mathbb{R}^k \setminus (A_1 \cup \dots \cup A_m)) = 0$ ) and  $g$  is one-to-one on  $A_j$  w/ non-vanishing Jacobian (i.e.  $\det(\partial g(x)/\partial x) \neq 0$  on  $A_j$ ):

$$\Rightarrow f_Y(y) = \sum_{j=1}^m |\det(\partial h_j(y)/\partial y)| f_X(h_j(y)), \quad h_j(y) = g^{-1}(y) \text{ on } A_j$$

Cochran's theorem:  $X \sim N_n(\mu, I_n)$ ;  $X^T X = X^T A_1 X + \dots + X^T A_k X$  ( $A_j$  is  $n \times n$  symmetric w/ rank  $n_j$ ;  $X^T A_j X$  have non-central chi-square distribution  $\chi^2_{n_j}(\delta_j)$  and are ~~independent~~ independent iff  $n = \sum_{j=1}^k n_j \rightarrow \delta_j = \mu^T A_j \mu$  &  $\mu^T \mu = \sum_{j=1}^k \delta_j$

Moments and moment inequalities: Def.:  $EX = \int_{\Omega} X dP$  for random variable  $X$  on  $(\Omega, \mathcal{F}, P)$

if  $EX^k$  finite ( $k$  positive integer)  $\rightarrow k$ th moment of  $X$  or  $P_X$

if  $E|X|^a < \infty$  ( $a \in \mathbb{R}$ )  $\rightarrow a$ th absolute moment

if  $E|X|^a < \infty$  ( $a \in \mathbb{R}_{>0}$ )  $\rightarrow EX^k$  finite  $\forall k \leq a$

if  $\mu = EX$  &  $E(X-\mu)^k$  finite  $\rightarrow k$ th central moment ( $k=2$  called variance)

For vectors  $X = (X_1, \dots, X_k)$ :  $EX = (EX_1, \dots, EX_k)$

variance-covariance matrix:  $\text{Var}(X) = E((X-EX)(X-EX)^T)$

$$Y = c^T X \quad (c \in \mathbb{R}^k) \Rightarrow EY = c^T EX, \quad \text{Var}(Y) = c^T \text{Var}(X) c$$