

FORMULA SHEET EXAM 2

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$$\text{Var}(X) = \sigma^2 = E[(X - \mu)^2] \quad \mu = E(X). \quad (1)$$

$$\text{Cov}(X, Y) = E[(X - \mu_X)(Y - \mu_Y)] \quad (2)$$

$$\rho(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y} \quad (3)$$

$$\text{Var}(X) = E[\text{Var}(X|Y)] + \text{Var}[E(X|Y)]. \quad (4)$$

$$P(X \geq t) \leq \frac{E(X)}{t}. \quad (5)$$

$$P(|X - E(X)| \geq t) \leq \frac{\text{Var}(X)}{t^2}. \quad (6)$$

$$f_{\mathbf{Y}}(y_1, y_2, \dots, y_m) = \sum_{\substack{(x_1, \dots, x_n) : r_i(x_1, \dots, x_n) = y_i \\ \forall i=1..m}} f_{\mathbf{X}}(x_1, \dots, x_n) \quad (7)$$

$$f_{\mathbf{Y}}(y_1, y_2, \dots, y_n) = \begin{cases} f_{\mathbf{X}}(s_1(), s_2(), \dots, s_n()) |J|, & \text{for } (y_1, y_2, \dots, y_n) \in \mathcal{Y} \subseteq R^n; \\ 0, & \text{otherwise.} \end{cases} \quad (8)$$

where,

$$\begin{array}{lll} Y_1 = r_1(X_1, \dots, X_n) & & X_1 = s_1(Y_1, \dots, Y_n) \\ Y_2 = r_2(X_1, \dots, X_n) & \text{unique} & X_2 = s_2(Y_1, \dots, Y_n) \\ \vdots & \text{transformation} & \vdots \\ Y_n = r_n(X_1, \dots, X_n) & \longrightarrow & X_n = s_n(Y_1, \dots, Y_n); \end{array} \quad (9)$$

and

$$J = \det \begin{bmatrix} \frac{\partial s_1}{\partial y_1} & \cdots & \frac{\partial s_1}{\partial y_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial s_n}{\partial y_1} & \cdots & \frac{\partial s_n}{\partial y_n} \end{bmatrix} \quad (\text{Jacobian}); \quad (10)$$

and

- \mathcal{X} is the support of $X_1, \dots, X_n : \mathcal{X} = \{\mathbf{x} : f_{\mathbf{x}}(\mathbf{x}) > 0\}$.
- \mathcal{Y} is the induced support of $Y_1, \dots, Y_n : \mathcal{Y} = \{\mathbf{y} : \mathbf{y} = r(\mathbf{x}) \text{ for all } \mathbf{x} \in \mathcal{X}\}$.
- $(x_1, \dots, x_n) \in \mathcal{X} \iff (y_1, \dots, y_n) \in \mathcal{Y}$. (11)

If $X_i \sim N(\mu_i, \sigma_i^2)$ and all n X_i are mutually independent, then

$$H = \sum_{i=0}^n a_i X_i + b_i \sim N\left(\sum_{i=0}^n a_i \mu_i + b_i, \sum_{i=0}^n a_i^2 \sigma_i^2\right). \quad (12)$$

If $Y \sim N(0, 1)$, then

$$Z = Y^2 \sim \chi^2_{(1)} \quad (13)$$

If $X_1 \sim \chi^2_{(p)}$ and $X_2 \sim \chi^2_{(q)}$ are independent, then

$$H = X_1 + X_2 \sim \chi^2_{(p+q)} \quad (14)$$