

|| PROBLEM SET - 1 ||

17 $AB = I \Rightarrow \text{rank}(AB) = \text{rank}(I) = n$

~~Now~~ Now, $\text{rank}(AB) \leq \min\{\text{rank}(A), \text{rank}(B)\} \leq n$
 $= n$.

~~Rank~~

[as A & B are $n \times n$ matrices]

$\therefore \text{rank}(A) = n = \text{rank}(B)$.

$\therefore A$ & B are invertible.

$AB = I \Rightarrow B = A^{-1}$

$\therefore BA = A^{-1} \cdot A = I$.

27 Given: $A^2 = A$

Let say, $\text{rank}(A) = n$.

$A = B^{n \times n} C^{n \times n}$ where, B has left inverse & C has right inverse.

$A^2 = A \cdot A = B C \cdot B C = B C$

$\Rightarrow C B C = C$ [Pre Multiplying with left inverse of B]

$\Rightarrow C B = I_n$ [Post Multiplying with right inverse of C]

$\text{trace}(A) = \text{trace}(B C) = \text{trace}(C B) = \text{trace}(I_n) = n$.

$$3) X = (x_1, x_2, x_3)' \quad \text{Var}(X) = \begin{pmatrix} 5 & 2 & 3 \\ 2 & 3 & 0 \\ 3 & 0 & 3 \end{pmatrix}$$

$$\textcircled{a} \quad x_1 - 2x_2 + x_3 = (1 \quad -2 \quad 1)' X \\ = L'X$$

$$\text{Var}(L'X) = L' \text{Var}(X) L$$

$$= (1 \quad -2 \quad 1)' \begin{pmatrix} 5 & 2 & 3 \\ 2 & 3 & 0 \\ 3 & 0 & 3 \end{pmatrix} \begin{pmatrix} 1 \\ -2 \\ 1 \end{pmatrix}$$

$$= (1 \quad -2 \quad 1) \begin{pmatrix} 4 \\ -4 \\ 6 \end{pmatrix}$$

$$= 18$$

$$\textcircled{b} \quad Y = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} x_1 + x_2 \\ x_1 + x_2 + x_3 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \\ = AX$$

$$\text{Var}(Y) = \text{Var}(AX)$$

$$= A \text{Var}(X) A'$$

$$= \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 5 & 2 & 3 \\ 2 & 3 & 0 \\ 3 & 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 0 & 1 \end{bmatrix}$$

$$\begin{aligned}
 47 \quad \bar{x} &= \frac{1}{n} \sum_{i=1}^n x_i = \left[\frac{1}{n} \quad \frac{1}{n} \quad \frac{1}{n} \quad \dots \quad \frac{1}{n} \right] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \\
 &= \underline{\underline{\mathbf{1}}}' \underline{\underline{\mathbf{x}}} \quad \left[\underline{\underline{\mathbf{1}}} = \left(\frac{1}{n} \quad \frac{1}{n} \quad \dots \quad \frac{1}{n} \right)' \right. \\
 &\quad \left. \underline{\underline{\mathbf{x}}} = (x_1, x_2, \dots, x_n)' \right]
 \end{aligned}$$

$$\begin{aligned}
 Q &= \sum_{i=1}^n (x_i - \bar{x})^2 = \sum_{i=1}^n x_i^2 - n \bar{x}^2 \\
 &= \sum_{i=1}^n x_i \cdot x_i - \frac{1}{n} (\sum x_i) (\sum x_i) \\
 &= \sum_{i=1}^n x_i^2 - \frac{1}{n} \sum_{\substack{i \neq j \\ i, j}} x_i x_j \\
 &= \sum_{i=1}^n \left(1 - \frac{1}{n} \right) x_i^2 - \frac{1}{n} \sum_{\substack{i \neq j \\ i, j}} x_i x_j \\
 &= \underline{\underline{\mathbf{x}}}^T \mathbf{A} \underline{\underline{\mathbf{x}}}
 \end{aligned}$$

where $\mathbf{A} = \begin{bmatrix} 1 - \frac{1}{n} & -\frac{1}{n} & -\frac{1}{n} & \dots & -\frac{1}{n} \\ -\frac{1}{n} & 1 - \frac{1}{n} & -\frac{1}{n} & \dots & -\frac{1}{n} \\ & & \vdots & & \\ -\frac{1}{n} & -\frac{1}{n} & -\frac{1}{n} & \dots & 1 - \frac{1}{n} \end{bmatrix}$

A is symmetric

$$= \mathbf{I} - \frac{1}{n} \underline{\underline{\mathbf{1}}} \underline{\underline{\mathbf{1}}}^T$$

$$\mathbf{A}^2 = \left(\mathbf{I} - \frac{1}{n} \underline{\underline{\mathbf{1}}} \underline{\underline{\mathbf{1}}}^T \right) \left(\mathbf{I} - \frac{1}{n} \underline{\underline{\mathbf{1}}} \underline{\underline{\mathbf{1}}}^T \right)$$

$$= \left(\mathbf{I} - \frac{1}{n} \underline{\underline{\mathbf{1}}} \underline{\underline{\mathbf{1}}}^T - \frac{1}{n} \underline{\underline{\mathbf{1}}} \underline{\underline{\mathbf{1}}}^T + \frac{1}{n^2} \cdot n \underline{\underline{\mathbf{1}}} \underline{\underline{\mathbf{1}}}^T \right)$$

$$= \mathbf{I} - \frac{1}{n} \underline{\underline{\mathbf{1}}} \underline{\underline{\mathbf{1}}}^T$$

$$= \mathbf{A}$$

$$A \sim = \begin{bmatrix} 1 & -1 & 0 & \dots & 0 \\ -1 & 2 & -1 & \dots & 0 \\ 0 & -1 & 2 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 2 & -1 \\ 0 & 0 & 0 & \dots & -1 & 1 \end{bmatrix}$$

$$= \frac{1}{n} \mathbf{1} \mathbf{1}^T - \frac{1}{n^2} \cdot \mathbf{1} \cdot \mathbf{1}^T$$

$$= 0$$

$\Rightarrow \bar{x}$ & $\sum (x_i - \bar{x})^2$ are independent

57 $x_1, \dots, x_n \stackrel{i.i.d.}{\sim} N(\mu, \sigma^2)$

$$\bar{x} = \mathbf{e}' \mathbf{x} \quad \mathbf{e}' = \left(\frac{1}{n} \quad \frac{1}{n} \dots \frac{1}{n} \right)'$$

$$\sum_{i=1}^{n-1} (x_i - x_{i+1})^2$$

$$= (x_1 - x_2)^2 + (x_2 - x_3)^2 + \dots + (x_{n-1} - x_n)^2$$

$$= x_1^2 + 2x_2^2 + \dots + 2x_{n-1}^2 + x_n^2 - 2x_1x_2 - 2x_2x_3 - \dots - 2x_{n-1}x_n$$

$$= \mathbf{x}^T A \mathbf{x}$$

$$A = \begin{bmatrix} 1 & -1 & 0 & 0 & \dots & 0 & 0 \\ -1 & 2 & -1 & 0 & \dots & 0 & 0 \\ 0 & -1 & 2 & -1 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \dots & 2 & -1 \\ 0 & 0 & 0 & 0 & \dots & -1 & 1 \end{bmatrix}$$

A is symmetric.

$$A^2 = \begin{bmatrix} 1 & -1 & 0 & 0 & \dots & 0 & 0 \\ -1 & 2 & -1 & 0 & \dots & 0 & 0 \\ 0 & -1 & 2 & -1 & \dots & 0 & 0 \\ 0 & 0 & -1 & 2 & \dots & 0 & 0 \\ & & & \vdots & & & \\ 0 & 0 & 0 & 0 & \dots & 2 & -1 \\ 0 & 0 & 0 & 0 & \dots & -1 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & -1 & 0 & 0 & \dots & 0 & 0 \\ -1 & 2 & -1 & 0 & \dots & 0 & 0 \\ 0 & -1 & 2 & -1 & \dots & 0 & 0 \\ & & & \vdots & & & \\ 0 & 0 & 0 & 0 & \dots & 2 & -1 \\ 0 & 0 & 0 & 0 & \dots & -1 & 1 \end{bmatrix}$$

=

$$b) \quad Y \sim N_2(0, \Sigma) \quad \Sigma = (\sigma_{ij})$$

$$\Sigma = \begin{pmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{pmatrix}$$

$$\sigma_{12} = \sigma_{21}$$

$$Y' \Sigma^{-1} Y = Y_1^2 / \sigma_{11}$$

$$= (Y_1 \ Y_2) \begin{pmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{12} & \sigma_{22} \end{pmatrix}^{-1} \begin{pmatrix} Y_1 \\ Y_2 \end{pmatrix} = Y_1^2 / \sigma_{11}$$

$$= (Y_1 \ Y_2) \frac{1}{\Delta} \begin{pmatrix} \sigma_{22} & -\sigma_{12} \\ -\sigma_{12} & \sigma_{11} \end{pmatrix} \begin{pmatrix} Y_1 \\ Y_2 \end{pmatrix} = \frac{Y_1^2}{\sigma_{11}} \left[\Delta = \sigma_{11}\sigma_{22} - \sigma_{12}^2 \right]$$

$$= Y_1^2 \left(\frac{\sigma_{22}}{\sigma_{11}\sigma_{22} - \sigma_{12}^2} - \frac{1}{\sigma_{11}} \right) + \frac{\sigma_{11}}{\sigma_{11}\sigma_{22} - \sigma_{12}^2} Y_2^2$$

$$- \frac{2\sigma_{12}}{\sigma_{11}\sigma_{22} - \sigma_{12}^2} Y_1 Y_2$$

$$= Y_1^2 \left(\frac{\cancel{\sigma_{11}}\sigma_{22} - \cancel{\sigma_{11}}\sigma_{22} + \sigma_{12}^2}{\sigma_{11}(\sigma_{11}\sigma_{22} - \sigma_{12}^2)} \right) + \frac{\sigma_{11}}{\sigma_{11}\sigma_{22} - \sigma_{12}^2} Y_2^2$$

$$- \frac{2\sigma_{12}}{\sigma_{11}\sigma_{22} - \sigma_{12}^2} Y_1 Y_2$$

$$A = \frac{1}{\Delta} \begin{bmatrix} \frac{\sigma_{12}^2}{\sigma_{11}} & -\sigma_{12} \\ -\sigma_{12} & \sigma_{11} \end{bmatrix}$$

$$A \Sigma = \frac{1}{\Delta} \begin{bmatrix} \sigma_{12}^2 / \sigma_{11} & -\sigma_{12} \\ -\sigma_{12} & \sigma_{11} \end{bmatrix} \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{12} & \sigma_{22} \end{bmatrix}$$

$$= \frac{1}{\Delta} \begin{bmatrix} 0 & \frac{\sigma_{12}^3}{\sigma_{11}} - \sigma_{12}\sigma_{11} \\ 0 & -\sigma_{12}^2 + \sigma_{11}\sigma_{22} \end{bmatrix} = \begin{bmatrix} 0 & -\sigma_{12}/\sigma_{11} \\ 0 & 1 \end{bmatrix}$$

$$\begin{aligned}
 (A\Sigma)^2 &= \begin{bmatrix} 0 & -\sigma_{12}/\sigma_{11} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & -\sigma_{12}/\sigma_{11} \\ 0 & 1 \end{bmatrix} \\
 &= \begin{bmatrix} 0 & -\sigma_{12}/\sigma_{11} \\ 0 & 1 \end{bmatrix} = A\Sigma
 \end{aligned}$$

$\therefore A\Sigma$ is idempotent.

Now, $Y \sim N(0, \Sigma)$

$\Delta A\Sigma$ is idempotent.

$$Y'AY \sim \chi^2_{\text{rank}(A), \lambda}$$

$$\begin{aligned}
 \text{Now, rank}(A) &= 1 & \left[\det(A) = 0 \right. \\
 \text{and } \lambda &= \mu^T \Delta \mu & \left. \therefore \text{rank}(A) < 2 \right.
 \end{aligned}$$

$$\begin{aligned}
 &= 0 \text{ (if } \mu=0) \Rightarrow \text{rank}(A) = 1 \quad [A \neq 0] \\
 \therefore Y'AY &\sim \chi^2_1 \text{ [Proved]}
 \end{aligned}$$

$$7) \text{cov}(\hat{\beta}_0, \hat{\beta}_1)$$

$$\text{var}(\hat{\beta}_1) = \frac{\sigma^2}{S_{xx}}$$

$$= \text{cov}(\bar{y} - \hat{\beta}_1 \bar{x}, \hat{\beta}_1)$$

$$= \text{cov}(\bar{y}, \hat{\beta}_1) - \bar{x} \text{cov}(\hat{\beta}_1, \hat{\beta}_1)$$

$$= \text{cov}(\bar{y}, \hat{\beta}_1) - \bar{x} \text{var}(\hat{\beta}_1)$$

$$\text{cov}(\bar{y}, \hat{\beta}_1) = \frac{1}{n} \text{cov}\left(\sum_{i=1}^n y_i, \sum_{i=1}^n c_i y_i\right) \quad c_i = \frac{x_i - \bar{x}}{\sum (x_i - \bar{x})^2}$$

$$= \frac{1}{n} \sum_{i=1}^n \sum_{j=1}^n c_j \text{cov}(y_i, y_j)$$

$$= \frac{1}{n} \sum_{i=1}^n c_i \text{var}(y_i) \quad [\text{cov}(y_i, y_j) = 0 \text{ for } i \neq j]$$

$$= \frac{1}{n} \sum_{i=1}^n \frac{x_i - \bar{x}}{\sum (x_i - \bar{x})^2} \cdot \sigma^2$$

$$= \frac{\sigma^2}{n} \cdot 0 = 0$$

$$\text{cov}(\hat{\beta}_0, \hat{\beta}_1) = \cancel{\text{cov}} - \bar{x} \cdot \text{var}(\hat{\beta}_1)$$

$$= -\bar{x} \cdot \frac{\sigma^2}{S_{xx}} = 0$$

$$\Leftrightarrow \bar{x} = 0 \quad (\text{proven})$$

$$8). y_i = \beta_0 + \beta_1 (x_{i1} - \bar{x}_1) + \dots + \beta_{p-1} (x_{ip-1} - \bar{x}_{p-1}) + \epsilon_i$$

$$\underline{y} = \beta_0 \underline{1} + \tilde{X} \underline{\beta} + \underline{\epsilon}$$

$$\underline{\epsilon} = (\underline{y} - \beta_0 \underline{1} - \tilde{X} \underline{\beta})$$

$$S = \underline{\epsilon}^T \underline{\epsilon} = (\underline{y} - \beta_0 \underline{1} - \tilde{X} \underline{\beta})^T (\underline{y} - \beta_0 \underline{1} - \tilde{X} \underline{\beta})$$

$$\begin{aligned} &= \underline{y}^T \underline{y} - \beta_0 \underline{1}^T \underline{y} - \underline{\beta}^T \tilde{X}^T \underline{y} \\ &\quad - \beta_0 \underline{y}^T \underline{1} + \beta_0^2 \underline{1}^T \underline{1} + \underline{\beta}^T \tilde{X}^T \beta_0 \underline{1} \\ &\quad - \cancel{\underline{\beta}^T \tilde{X}^T \underline{y}} \underline{y}^T \tilde{X} \underline{\beta} + \beta_0 \underline{1}^T \tilde{X} \underline{\beta} \\ &\quad + \underline{\beta}^T \tilde{X}^T \tilde{X} \underline{\beta} \end{aligned}$$

$$\frac{\partial S}{\partial \beta_0} = - \underline{1}^T \underline{y} - \underline{y}^T \underline{1} + 2\beta_0 \underline{1}^T \underline{1} + \underline{\beta}^T \tilde{X}^T \underline{1} + \underline{1}^T \tilde{X} \underline{\beta}$$

$$= -2\underline{y}^T \underline{1} + 2n\beta_0 + 2(\tilde{X}\underline{\beta})^T \underline{1} = 0$$

$$\Rightarrow \hat{\beta}_0 = \frac{1}{n} [\underline{y}^T \underline{1} - (\tilde{X}\underline{\beta})^T \underline{1}]$$

$$= \cancel{\bar{y}} \quad \left[\text{Since, column sum of } \tilde{X} \text{ is } 0 \right]$$

$$\left(\frac{\partial}{\partial x} a'x = a \right)$$

$$\frac{\partial S}{\partial \underline{\beta}} = -2\tilde{X}^T \underline{y} + 2\beta_0 \tilde{X}^T \underline{1} + 2\tilde{X}^T \tilde{X} \underline{\beta} = 0$$

$$\Rightarrow -2\tilde{X}^T \underline{y} + \cancel{2\tilde{X}^T \underline{1} \beta_0} 2\tilde{X}^T \underline{\beta} = 0 \quad [\tilde{X}^T \underline{1} = 0]$$

$$\Rightarrow \hat{\underline{\beta}} = (\tilde{X}^T \tilde{X})^{-1} \tilde{X}^T \underline{y}$$

$$97. y_i = \beta_0 + \beta_1 (x_{1i} - \bar{x}_1)/s_1 + \dots + \beta_{p-1} (x_{(p-1)i} - \bar{x}_{p-1})/s_{p-1} + \epsilon_i$$

$$\tilde{y} = \beta_0 \tilde{1} + \tilde{X} \tilde{\beta} + \tilde{\epsilon}$$

$$\tilde{X} = \begin{bmatrix} \frac{x_{11} - \bar{x}_1}{s_1} & \dots & \frac{x_{(p-1)1} - \bar{x}_{p-1}}{s_{p-1}} \\ \vdots & & \vdots \\ \frac{x_{1n} - \bar{x}_1}{s_1} & \dots & \frac{x_{(p-1)n} - \bar{x}_{p-1}}{s_{p-1}} \end{bmatrix}$$

$$S = \tilde{\epsilon}^T \tilde{\epsilon} = (\tilde{y} - \beta_0 \tilde{1} - \tilde{X} \tilde{\beta})^T (\tilde{y} - \beta_0 \tilde{1} - \tilde{X} \tilde{\beta})$$

$$\frac{\partial S}{\partial \beta_0} = 0 \Rightarrow \hat{\beta}_0 = \frac{1}{n} [\tilde{y}^T \tilde{1} - (\tilde{X} \tilde{\beta})^T \tilde{1}]$$

$$= \bar{y} - \frac{1}{n} (\tilde{X} \hat{\beta})^T \tilde{1} = \bar{y}$$

$$\frac{\partial S}{\partial \beta} = 0 \Rightarrow -\tilde{X}^T \tilde{y} + \beta_0 \tilde{X}^T \tilde{1} + \tilde{X}^T \tilde{X} \beta = 0$$

$$\Rightarrow \hat{\beta} = (\tilde{X}^T \tilde{X})^{-1} \tilde{X}^T \tilde{y}$$

$$\Rightarrow -\tilde{X}^T \tilde{y} + \left(\frac{1}{n} \tilde{y}^T \tilde{1} - \frac{1}{n} (\tilde{X} \hat{\beta})^T \tilde{1} \right) \tilde{X}^T \tilde{1} + \tilde{X}^T \tilde{X} \hat{\beta} = 0$$

$$\Rightarrow -\tilde{X}^T \tilde{y} + \frac{1}{n} \tilde{1}^T \tilde{y} \cdot \tilde{X}^T \tilde{1} - \frac{1}{n} \beta^T \tilde{X}^T \tilde{1} + \tilde{X}^T \tilde{X} \hat{\beta} = 0$$

$$\Rightarrow -\tilde{X}^T \tilde{y} + \frac{1}{n} \tilde{1}^T \tilde{y} \cdot \tilde{X}^T \tilde{1} - \frac{1}{n} \beta^T \tilde{X}^T \tilde{1} + \tilde{X}^T \tilde{X} \hat{\beta} = 0$$

$$\text{var}(\hat{\beta}) = \sigma^2 (\tilde{x}^T \tilde{x})^{-1}$$

107.

$$Y_1 = \alpha_1 + \epsilon_1$$

$$Y_2 = 2\alpha_1 - \alpha_2 + \epsilon_2$$

$$Y_3 = \alpha_1 + 2\alpha_2 + \epsilon_3$$

$$\underline{y} = \begin{bmatrix} 1 & 0 \\ 2 & -1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \end{bmatrix} + \underline{\epsilon}$$

$$= X\underline{\beta} + \underline{\epsilon}$$

$$\hat{\beta} = (X^T X)^{-1} X^T \underline{y}$$

$$X^T X = \begin{bmatrix} 1 & 2 & 1 \\ 0 & -1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 2 & -1 \\ 1 & 2 \end{bmatrix} \\ = \begin{bmatrix} 6 & 0 \\ 0 & 5 \end{bmatrix}$$

$$(X^T X)^{-1} = \begin{bmatrix} 1/6 & 0 \\ 0 & 1/5 \end{bmatrix}$$

$$\hat{\beta} = (X^T X)^{-1} X^T \underline{y} = \begin{pmatrix} 1/6 & 0 \\ 0 & 1/5 \end{pmatrix} \begin{pmatrix} 1 & 2 & 1 \\ 0 & -1 & 2 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix}$$

$$= \begin{pmatrix} 1/6 & 1/3 & 1/6 \\ 0 & -1/5 & 2/5 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix}$$

$$\begin{pmatrix} \hat{\alpha}_1 \\ \hat{\alpha}_2 \end{pmatrix} = \begin{pmatrix} 1/6 y_1 + 1/3 y_2 + 1/6 y_3 \\ -1/5 y_2 + 2/5 y_3 \end{pmatrix}$$

$$H_0: \alpha_1 = \alpha_2 \Rightarrow \alpha_1 - \alpha_2 = 0$$

~~$$A = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \hat{\beta} = \begin{pmatrix} \alpha_1 \\ \alpha_2 \end{pmatrix}, c = 0$$~~

$$A = [1 \ -1], \hat{\beta} = \begin{pmatrix} \alpha_1 \\ \alpha_2 \end{pmatrix}, c = 0$$

$$\hat{\beta}_R = \hat{\beta} + (x'x)^{-1} A' [A(x'x)^{-1} A']^{-1} [c - A\hat{\beta}]$$

~~$$= \begin{pmatrix} 1/6 y_1 + 1/3 y_2 + 1/6 y_3 \\ -1/5 y_2 \end{pmatrix}$$~~

$$= \hat{\beta} + \begin{bmatrix} 1/6 & 0 \\ 0 & 1/5 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \left[\begin{bmatrix} 1 & -1 \\ 1/6 & 0 \\ 0 & 1/5 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right]^{-1} (0 - [1 \ -1] \hat{\beta})$$

$$\hat{\sigma}_R^2 = \frac{\|y - x\hat{\beta}_R\|^2}{n}$$

$$\hat{\sigma}^2 = \frac{\|y - x\hat{\beta}\|^2}{n}$$

$$\Delta = \left(\frac{\hat{\sigma}^2}{\hat{\sigma}_R^2} \right)^{n/2}$$

LR.

$$\text{Test statistic} = \frac{(3-2)}{1} \left(\frac{\Delta^{2/n}}{\Delta^{2/n} - 1} \right) \left(\Delta^{2/n} - 1 \right)$$