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Exam in TMA4267 Linear Statistical Models Tuesday June 2, 2009 Time: 09:00-13:00

Permitted assisting material: None

You may answer in English or Norwegian. Du kan besvare enten på engelsk eller norsk.

Notation: $y \sim \mathcal{N}_p(\mu, \Sigma)$, denotes that $y = (y_1, y_2, \dots, y_p)^T$ is normal distributed with mean μ and covariance matrix Σ . I is the diagonal matrix.

Problem 1 Let

$$m{y} \, \sim \, \mathcal{N}_2 \left(\left(egin{matrix} 1 \ 1 \end{matrix}
ight), \left(egin{matrix} 9 & 2 \ 2 & 9 \end{matrix}
ight)
ight)$$

a) What is the distribution of $z_1 = y_1 + y_2$? What is the distribution of $z_2 = z_1 - 2y_2$? What is $Cov(z_1, z_2)$?

Problem 2 Assume the data y follows the model

$$y = X_1 \beta_1 + X_2 \beta_2 + \epsilon$$

where $\epsilon \sim \mathcal{N}_n(0, \sigma^2 I)$, but you estimate the following model using ordinary least squares,

$$y = X_1 \beta_1^* + \epsilon^*.$$

Here, $X_1 \neq X_2$, and both X_1 and X_2 have full rank.

a) Show that $\widehat{\boldsymbol{\beta}}_1^*$ has the following properties

$$\mathbf{E}\left(\widehat{\boldsymbol{\beta}}_{1}^{*}\right) = \boldsymbol{\beta}_{1} + \left(\boldsymbol{X}_{1}^{T}\boldsymbol{X}_{1}\right)^{-1}\boldsymbol{X}_{1}^{T}\boldsymbol{X}_{2}\boldsymbol{\beta}_{2}$$

and

$$\operatorname{Cov}\left(\widehat{\boldsymbol{eta}}_{1}^{*}\right)=\sigma^{2}\left(\boldsymbol{X}_{1}^{T}\boldsymbol{X}_{1}\right)^{-1}.$$

b) Under what condition (of X_1 and X_2) is $\widehat{\beta}_1^*$ unbiased? What is the geometrical interpretation of this condition?

Problem 3 The Cook's distance is defined in the book as

$$D_i = \frac{\left(\widehat{\boldsymbol{y}}_{(i)} - \widehat{\boldsymbol{y}}\right)^T \left(\widehat{\boldsymbol{y}}_{(i)} - \widehat{\boldsymbol{y}}\right)}{(k+1)s^2}$$

a) Explain each term in the Cook's distance.Explain the use of the Cook's distance.

Problem 4 Let

$$y_1, y_2, \ldots, y_n \stackrel{\text{iid}}{\sim} \mathcal{N}_1(\mu, \sigma^2).$$

Define

$$m{j} = egin{pmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix} \quad ext{and} \quad m{J} = m{j} m{j}^T = egin{pmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{pmatrix}$$

and $\overline{y} = \frac{1}{n} \sum_{i=1}^{n} y_i$.

a) Show that

$$\sum_{i=1}^{n} y_i^2 = \sum_{i=1}^{n} (y_i - \overline{y})^2 + n\overline{y}^2$$

can be expressed as

$$\boldsymbol{y}^T \boldsymbol{I} \boldsymbol{y} = \boldsymbol{y}^T \left(\boldsymbol{I} - \frac{1}{n} \boldsymbol{J} \right) \boldsymbol{y} + \boldsymbol{y}^T \left(\frac{1}{n} \boldsymbol{J} \right) \boldsymbol{y}.$$

b) Show that

1. $I - \frac{1}{n}J$ and $\frac{1}{n}J$ are idempotent.

2.
$$\left(I-\frac{1}{n}J\right)\left(\frac{1}{n}J\right)=0$$

Problem 4 continuous on the next page...

Here is a transcript of two main results from the book:

Theorem 5.5. Let y be distributed as $\mathcal{N}_p(\mu, \Sigma)$, let A be a symmetric matrix of constants of rank r, and let $\lambda = \frac{1}{2}\mu^T A\mu$. Then $y^T Ay$ is $\chi^2(r, \lambda)$, if and only if $A\Sigma$ is idempotent.

Theorem 5.6a. Suppose B is a $k \times p$ matrix of constants, A is a $p \times p$ symmetric matrix of constants, and y is distributed as $\mathcal{N}_p(\mu, \Sigma)$. Then By and y^TAy are independent if and only if $B\Sigma A = 0$.

c) Use Theorem 5.5 and Theorem 5.6a to show that for $\mu = 0$, then

1.
$$\frac{\sum_{i=1}^{n} (y_i - \overline{y})^2}{\sigma^2} \sim \chi_{n-1}^2$$

2.
$$n \frac{\overline{y}^2}{\sigma^2} \sim \chi_1^2$$

3.
$$\frac{\sum_{i=1}^{n}(y_i-\overline{y})^2}{\sigma^2} \quad \text{and} \quad n\frac{\overline{y}^2}{\sigma^2} \quad \text{are independent.}$$

d) Use the results in c) to construct a F-test (Fisher-test) for

$$H_0: \mu = 0, \qquad H_1: \mu \neq 0$$

with significance level α .

e) Show that the test in d) is equivalent to a (classical) student-t test for testing the same hypothesis.