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## High Dimensional Statistics: Exercise Sheet 2

The exercises that I ask to be corrected are: 1, 4

1. Show: If  $Z \sim \mathcal{N}_n(0, \Sigma)$  and  $\Sigma$  is invertible, then  $Z^T \Sigma^{-1} Z \sim \chi^2_{(n)}$ .

**Solution.** In general, we can decompose an arbitrary matrix  $P$  in the form  $P = U^{-1} D U$ , where  $D$  is the diagonal matrix containing eigenvalues of  $P$ , i.e.,  $D = \text{diag}(\lambda_1, \dots, \lambda_n)$ . If  $P$  is symmetric,  $U^{-1} = U^T$  and hence  $P = U^T D U$ .

We first prove that  $\Sigma$  is symmetric:

$$\forall i, j \in \{1, \dots, n\}, (\Sigma)_{ij} = \text{cov}(Z_i, Z_j) = \text{cov}(Z_j, Z_i) = (\Sigma)_{ji}$$

Therefore, we can decompose  $\Sigma^{-1}$  as  $\Sigma^{-1} = U^T D U$  and hence we can state the following:

$$\begin{aligned} Z^T \Sigma^{-1} Z &= Z^T (U^T D U) Z \\ &= (Z^T U^T) D (U Z) \\ &= (U Z)^T D (U Z) \end{aligned}$$

Now we note that since  $Z \sim \mathcal{N}_n(0, \Sigma)$  and  $U^T U = I$ , then  $Y = U Z \sim \mathcal{N}_n(0, \Sigma)$

Hence, we can state that

$$\begin{aligned} (U Z)^T D (U Z) &= Y^T D Y \\ &= Y^T \text{diag}(\lambda_1, \dots, \lambda_n) Y \\ &= \sum_{i=1}^n \lambda_i Y_i^2 \end{aligned}$$

Let  $A = \sum_{i=1}^n \lambda_i Y_i^2$ . We know that  $Y_i \sim \mathcal{N}_n(0, \sigma I_n)$ , and therefore  $\frac{Y_i}{\sigma} \sim \mathcal{N}_n(0, I_n)$ . Then,  $\frac{1}{\sigma^2} A = \frac{1}{\sigma^2} \sum_{i=1}^n \lambda_i Y_i^2$  will be summation of squares of independent normal random variables, which follow a Chi-squared distribution, i.e.,  $\frac{A}{\sigma^2} \sim \chi^2_{(n)}$ , and since  $\sigma^2$  is a constant, we can deduce that  $A = \sum_{i=1}^n \lambda_i Y_i^2 \sim \chi^2_{(n)}$