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## Statistical Modelling Exam 23/01/2025

Instructions for the exam:

- Notation: bold symbols indicate vectors,
- The Gaussian distribution is denoted as  $N(\mu, \sigma^2)$ , with  $\mu$  mean parameter and  $\sigma^2$  variance.
- When performing statistical tests, explicitly write: the system of hypotheses, test statistic and its distribution, observed value, reject region and conclusion.

### Exercise 1

On  $n = 20$  statistical units we observe the values of two continuous numeric variables  $(y_i, x_i)$ ,  $i = 1, \dots, n$ . To these data, it is fitted the linear regression model

$$Y_i = \beta_1 + \beta_2(x_i - \bar{x}) + \beta_3(x_i - \bar{x})^2 + \varepsilon_i$$

with  $\bar{x} = (1/20) \sum_{i=1}^{20} x_i$ , and  $(\varepsilon_1, \dots, \varepsilon_{20})$  independent random variables with Gaussian distribution  $N(0, 4)$ .

Answer the following questions:

- Write the parameter and sample space.
- Express the model in matrix form:  $\mathbf{Y} = X\boldsymbol{\beta} + \boldsymbol{\varepsilon}$ , explicitly stating how  $\mathbf{Y}$ ,  $X$ ,  $\boldsymbol{\beta}$ , and  $\boldsymbol{\varepsilon}$  are defined and their dimensions. Write the distribution of  $\mathbf{Y}$  and  $\boldsymbol{\varepsilon}$ .
- Write the likelihood and log-likelihood for the parameters of the model.
- Knowing that,

$$(X^T X)^{-1} = \begin{bmatrix} 0.8 & -1.9 & 2.5 \\ -1.9 & 5.9 & -9.4 \\ 2.5 & -9.4 & 18.7 \end{bmatrix}, \quad X^T \mathbf{y} = \begin{bmatrix} 21 \\ 14 \\ 4 \end{bmatrix}, \quad \mathbf{y}^T \mathbf{y} = 473.78,$$

and that the sample means of  $\mathbf{y}$  and  $\mathbf{x}$  are, respectively,  $\bar{y} = 0.2$  and  $\bar{x} = 8$ , obtain the maximum likelihood estimates of the regression parameters.

- Write the exact distribution of the estimator  $\hat{B}_2$  of  $\beta_2$ .
- Perform a test to evaluate whether it is reasonable to keep the quadratic term.
- Write the definition of the coefficient of determination  $R^2$ . The  $R^2$  of the fitted model is equal to 0.122, how do you interpret this value?
- Perform a test about the overall significance of the model using a 10% significance level.
- Let  $\mathbf{e} = \mathbf{y} - X^T \hat{\boldsymbol{\beta}}$  be the vector of the residuals. Indicate which of the following identities are true and motivate the answer:

$$\sum_{i=1}^{20} e_i = 0, \quad \sum_{i=1}^{20} e_i x_i = 0, \quad \sum_{i=1}^{20} e_i x_i = \bar{x} \bar{e}, \quad \sum_{i=1}^{20} e_i (x_i - \bar{x})^2 = 0.$$

## Exercise 2

The *Pima* dataset was collected by the National Institute of Diabetes and Digestive and Kidney Diseases. The objective of the dataset is to diagnostically predict whether or not a patient has diabetes, based on certain diagnostic measurements. In particular, the  $n = 724$  patients in this dataset are females at least 21 years old of Pima heritage.

The datasets has one response variable (**diabetes**: 1 = positive; 0 = negative), and it is known that, of these women, 249 have diabetes, while 475 do not.

Moreover, we have the following medical predictor variables:

- **pregnant**: Presence of present/past pregnancies: 0 = no pregnancies; 1 = at least one pregnancy.
- **glucose** : Plasma glucose concentration, numeric.
- **pressure**: Diastolic blood pressure (mm Hg), numeric.
- **BMI** : Body mass index, numeric.
- **age** : Age (years), numeric.

Fitting a logistic regression on R returns the following output (“model A”):

	Estimate	Std.Error	z value	Pr(> z )
(Intercept)	-8.9267	0.8537	-10.46	0.0000
<b>pregnant</b>	0.2465	0.2931	0.84	0.4004
<b>glucose</b>	0.0349	0.0035	9.92	0.0000
<b>pressure</b>	-0.0078	0.0084	-0.93	0.3515
<b>BMI</b>	0.0941	0.0154	6.09	0.0000
<b>age</b>	0.0328	0.0086	3.81	0.0001

Null deviance: 931.94 on 723 degrees of freedom

Residual deviance: 694.45 on 718 degrees of freedom

Answer the following:

- Write the corresponding theoretical model.
- Write the likelihood and log-likelihood functions for the regression parameters of the model.
- Provide the interpretation of the **age** and **pregnant** coefficients.
- Is it reasonable to remove the **pregnant** variable from the regression? Why?
- Define the concept of “odds” and how to interpret it.

A new model (“model B”) is then fitted removing the **pregnant** and **pressure** variables. This model returns the following output:

	Estimate	Std.Error	z value	Pr(> z )
(Intercept)	-9.0085	0.7261	-12.41	0.0000
<b>glucose</b>	0.0346	0.0035	9.90	0.0000
<b>BMI</b>	0.0884	0.0147	6.01	0.0000
<b>age</b>	0.0317	0.0079	3.99	0.0001

Null deviance: 931.94 on 723 degrees of freedom

Residual deviance: 696.15 on 720 degrees of freedom

- Perform a test to compare model A and model B using a 5% significance level. Which one do you prefer?
- According to model B, what is the probability of developing diabetes for a woman aged 25, with a glucose level equal to 99.75 and a BMI of 22?
- Define the null model. Obtain the estimate of the regression coefficients in this model.

	$p$						
	0.90	0.95	0.975	0.99	0.995	0.9975	0.999
$z_p$	1.2816	1.6449	1.9600	2.3263	2.5758	2.8070	3.0902

Table 1: Some quantiles of the Gaussian distribution:  $p = \mathbb{P}(Z \leq z_p)$ . Columns correspond to probabilities  $p$ .

	$p$						
	0.9	0.95	0.975	0.99	0.995	0.9975	0.999
$t_{2;p}$	1.8856	2.92	4.3027	6.9646	9.9248	14.089	22.3271
$t_{3;p}$	1.6377	2.3534	3.1824	4.5407	5.8409	7.4533	10.2145
$t_{17;p}$	1.3334	1.7396	2.1098	2.5669	2.8982	3.2224	3.6458
$t_{18;p}$	1.3304	1.7341	2.1009	2.5524	2.8784	3.1966	3.6105
$t_{19;p}$	1.3277	1.7291	2.093	2.5395	2.8609	3.1737	3.5794
$t_{20;p}$	1.3253	1.7247	2.086	2.528	2.8453	3.1534	3.5518

Table 2: Some quantiles of the t distribution:  $p = \mathbb{P}(T \leq t_{\alpha;p})$  with  $T \sim t_\alpha$ . Columns correspond to probabilities  $p$ . Rows correspond to different degrees of freedom  $\alpha$ .

	$p$						
	0.9000	0.9500	0.9750	0.9900	0.9950	0.9975	0.9990
$f_{1,18;p}$	3.0070	4.4139	5.9781	8.2854	10.2181	12.3208	15.3793
$f_{2,18;p}$	2.6239	3.5546	4.5597	6.0129	7.2148	8.5130	10.3899
$f_{3,18;p}$	2.4160	3.1599	3.9539	5.0919	6.0278	7.0351	8.4875
$f_{1,17;p}$	3.0262	4.4513	6.0420	8.3997	10.3842	12.5525	15.7222
$f_{2,17;p}$	2.6446	3.5915	4.6189	6.1121	7.3536	8.7006	10.6584
$f_{3,17;p}$	2.4374	3.1968	4.0112	5.1850	6.1556	7.2053	8.7269
$f_{1,16;p}$	3.0481	4.4940	6.1151	8.5310	10.5755	12.8201	16.1202
$f_{2,16;p}$	2.6682	3.6337	4.6867	6.2262	7.5138	8.9179	10.9710
$f_{3,16;p}$	2.4618	3.2389	4.0768	5.2922	6.3034	7.4027	9.0059

Table 3: Some quantiles of the F distribution:  $p = \mathbb{P}(F \leq f_{\alpha,\beta;p})$  with  $F \sim F_{\alpha,\beta}$ . Columns correspond to probabilities  $p$ . Rows correspond to different degrees of freedom  $\alpha$  and  $\beta$ .

	$p$						
	0.9	0.95	0.975	0.99	0.995	0.9975	0.999
$\chi_{2;p}^2$	4.6052	5.9915	7.3778	9.2103	10.5966	11.9829	13.8155
$\chi_{4;p}^2$	7.7794	9.4877	11.1433	13.2767	14.8603	16.4239	18.4668
$\chi_{6;p}^2$	10.6446	12.5916	14.4494	16.8119	18.5476	20.2494	22.4577

Table 4: Some quantiles of the  $\chi^2$  distribution:  $p = \mathbb{P}(\chi^2 \leq \chi_{\alpha;p}^2)$  with  $\chi^2 \sim \chi_\alpha^2$ . Columns correspond to probabilities  $p$ . Rows correspond to different degrees of freedom  $\alpha$ .