



Aufgabe 1 13 Punkte

For $k \in \mathbb{N}$ and $\alpha \in (0, 1)$, let Y be a discrete random variable with values in $M = \{1, \dots, k\}$ and corresponding probability mass function $p_Y(k, \alpha) : M \rightarrow [0, 1]$ given by

$$p_Y(k, \alpha) = \frac{(k-1)!}{(k-1)!} \cdot \frac{\alpha^{k-1}}{(1-\alpha)^k} \cdot (1-\alpha) \quad \text{for } k \in M.$$

It can be proven that the probability mass function for Exponential(1) applied on $k \in \mathbb{N}$ can be expressed in the following form:

$$p_Y(k, \alpha) = \exp\left(\frac{\ln(1-\alpha)}{\alpha} \cdot k\right) \cdot (1-\alpha) \quad \text{for } k \in M.$$

with $\alpha = \frac{1}{k+1}$ and $\ln(1-\alpha) = \ln\left(\frac{k}{k+1}\right) = \ln\left(\frac{k}{k+1}\right)$. For $k \in \mathbb{N}$, the corresponding natural parameter θ depending on α and an appropriate choice of η is

(a) For the following tasks, you need to use a representation of the form (**) and use it to solve the tasks (b)-(d).

(a) For the following tasks, you need to derive a representation of the form (**) for the probability mass function given by (1).

(i) For $\alpha = 1$, calculate the value of the corresponding natural parameter θ . 3 Punkte

Give the value of θ (rounded to three decimal places). 3 Punkte

-0.288

(ii) For $\alpha = 1$ and $k = 2$, calculate the expectation $E(Y)$. 2 Punkte

Give the value of $E(Y)$. 3 Punkte

3

(iii) For $\alpha = 1$ and $k = 2$, calculate the variance $\text{Var}(Y)$. 2 Punkte

Give the value of $\text{Var}(Y)$. 3 Punkte

2.6

(b) Consider three stochastically independent discrete random variables Y_1, Y_2, Y_3 , each having a probability mass function given by (1) for $k = 1$ and some parameters $\alpha_1, \alpha_2, \alpha_3 \in (0, 1)$, respectively, which are connected via the identity

$$\text{Var}(Y_1) = 8, \quad \text{Var}(Y_2) = 12, \quad \text{Var}(Y_3) = 2.$$

Further let $Y = (Y_1, Y_2, Y_3)^T$ and η a GLM given by $\eta(Y) = \eta(Y) = X\beta$ with canonical link function g , parameter vector $\beta = (\beta_1, \beta_2, \beta_3)^T \in \mathbb{R}^3$ and design matrix

$$X = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}.$$

In this framework and for the variances of Y_1, Y_2, Y_3 given above, determine the expected Fisher information matrix

$$I_Y(\beta) = \begin{pmatrix} I_{11} & I_{12} & I_{13} \\ I_{21} & I_{22} & I_{23} \\ I_{31} & I_{32} & I_{33} \end{pmatrix}$$

for the corresponding parameter vector β .

Hint: Part (a) can be solved independently of part (b), without deriving a representation of the form (**) for the given probability mass function.

Calculate the entries of $I_Y(\beta)$. 6 Punkte

Give the value of I_{11} . 2 Punkte

1/8

Give the value of I_{22} . 2 Punkte

1/12

a) i)

$$p_Y(k, \alpha) = \exp\left(\frac{\ln(1-\alpha)}{\alpha} \cdot k\right) \cdot (1-\alpha)$$

$$= \exp\left(\frac{\ln(1-\alpha)}{\alpha} \cdot k\right) \cdot \exp\left(\ln(1-\alpha)\right)$$

$$= \exp\left(\frac{\ln(1-\alpha)}{\alpha} \cdot k + \ln(1-\alpha)\right)$$

\rightarrow coefficient comparison:

$$\exp\left(\frac{\ln(1-\alpha)}{\alpha} \cdot k\right) = \exp\left(\frac{(k-1) \cdot \ln(1-\alpha)}{\alpha}\right)$$

$$\Rightarrow \frac{\exp(\theta)}{\exp(k\theta)} = \frac{\exp(\theta)}{\exp(k\theta)} = \left(\frac{\alpha-1}{\alpha}\right) \cdot \frac{1}{\alpha}$$

$$\Rightarrow \exp(\theta) = \frac{\alpha-1}{\alpha} = \frac{1}{\alpha} \quad \text{for } \alpha=1$$

$$\Rightarrow \theta = \ln\left(\frac{1}{1}\right) = 0$$

ii) 1.2.4

$$E(Y) = b'(\theta) = \frac{d}{d\theta} (\ln(\alpha))$$

$$\exp(\theta) = 1 - \frac{1}{\alpha}$$

$$1 - e^\theta = \frac{1}{\alpha} \Rightarrow \alpha = \frac{1}{1-e^\theta}$$

$$\Rightarrow E(Y) = \frac{d}{d\theta} \left(\ln\left(\frac{1}{1-e^\theta}\right) \right)$$

$$= \frac{d}{d\theta} \left(-\ln(1-e^\theta) \right)$$

$$= -\frac{1}{1-e^\theta} = -\frac{1}{1-e^0} = -1$$

iii) $\text{Var}(Y) = b''(\theta) \cdot a(\theta)$, $a(\theta) = \frac{1}{\alpha}$

$$= \frac{d}{d\theta} \left(\frac{e^\theta}{1-e^\theta} \right) = \frac{1}{2} \left(\frac{e^\theta}{1-e^\theta} + \frac{e^{2\theta}}{(1-e^\theta)^2} \right)$$

$$= \frac{e^\theta(1-e^\theta) + e^{2\theta}}{(1-e^\theta)^2} \cdot \frac{1}{2}$$

$$= \frac{e^\theta - e^{2\theta} + e^{2\theta}}{(1-e^\theta)^2} \cdot \frac{1}{2}$$

$$= \frac{e^\theta}{(1-e^\theta)^2} \cdot \frac{1}{2}$$

with $e^\theta = 3/4$

$$\Rightarrow \frac{3/4}{(3/4 - 1/4)^2} \cdot \frac{1}{2} = \frac{3/4}{1/8} \cdot \frac{1}{2} = 3 \cdot 2 = 6$$

b)

b) $b_1(\eta) = a \cdot b(\theta)$

$$= \frac{1}{k} \cdot b(\theta) \quad \text{with } k=1$$

$$\Rightarrow b_1(\eta) = b(\theta)$$

$$W = \begin{pmatrix} b'(\theta) & 0 & 0 \\ 0 & b'(\theta) & 0 \\ 0 & 0 & b'(\theta) \end{pmatrix}$$

$$\Rightarrow W = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$\Rightarrow Y = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$



11.2.22 Theorem (Log-Likelihood Derivative of GLM) [3]

Let $Y = (Y_1, \dots, Y_n)^T$ be a random vector with values in \mathbb{R}^n and let $X = (X_1, \dots, X_n)^T$ be a random vector with values in \mathbb{R}^n . Let $\eta = X\beta$ and let g be the canonical link function. Let $b(\eta) = (b_1(\eta), \dots, b_n(\eta))^T$ be the vector of natural parameters. Let $\beta = (\beta_1, \dots, \beta_n)^T$ be the vector of parameters. Let $W = (W_1, \dots, W_n)^T$ be the vector of weights. Let $I_Y(\beta)$ be the expected Fisher information matrix. Let $I_X(\beta)$ be the expected Fisher information matrix. Let $I(\beta)$ be the expected Fisher information matrix. Let $I(\beta)$ be the expected Fisher information matrix.

11.2.23 Definition (Observed Information Matrix)

Let $Y = (Y_1, \dots, Y_n)^T$ be a random vector with values in \mathbb{R}^n and let $X = (X_1, \dots, X_n)^T$ be a random vector with values in \mathbb{R}^n . Let $\eta = X\beta$ and let g be the canonical link function. Let $b(\eta) = (b_1(\eta), \dots, b_n(\eta))^T$ be the vector of natural parameters. Let $\beta = (\beta_1, \dots, \beta_n)^T$ be the vector of parameters. Let $W = (W_1, \dots, W_n)^T$ be the vector of weights. Let $I_Y(\beta)$ be the expected Fisher information matrix. Let $I_X(\beta)$ be the expected Fisher information matrix. Let $I(\beta)$ be the expected Fisher information matrix. Let $I(\beta)$ be the expected Fisher information matrix.

11.2.24 Information matrix for GLM with canonical link

For a GLM with canonical link function g , let $Y = (Y_1, \dots, Y_n)^T$ be a random vector with values in \mathbb{R}^n and let $X = (X_1, \dots, X_n)^T$ be a random vector with values in \mathbb{R}^n . Let $\eta = X\beta$ and let g be the canonical link function. Let $b(\eta) = (b_1(\eta), \dots, b_n(\eta))^T$ be the vector of natural parameters. Let $\beta = (\beta_1, \dots, \beta_n)^T$ be the vector of parameters. Let $W = (W_1, \dots, W_n)^T$ be the vector of weights. Let $I_Y(\beta)$ be the expected Fisher information matrix. Let $I_X(\beta)$ be the expected Fisher information matrix. Let $I(\beta)$ be the expected Fisher information matrix. Let $I(\beta)$ be the expected Fisher information matrix.

Hint: Part (a) can be solved independently of part (b), without doing a hypothesis test of the form $H_0: \theta = \theta_0$ vs $H_1: \theta > \theta_0$, for the given probability mass function.

Calculate the entries of $\mathbb{I}_Y(\theta)$. 6 Punkte

Give the value of $\hat{\theta}_{ML}$. 2 Punkte

Give the value of $\hat{\theta}_{LS}$. 2 Punkte

Give the value of $\hat{\theta}_{GLS}$. 2 Punkte

Aufgabe 2 7 Punkte

Let $\mu_1, \mu_2 \geq 0$ and $(X_n)_{n \geq 1}, (Y_n)_{n \geq 1}$ be two sequences of stochastically independent random variables with $X_n \sim \text{Pois}(\mu_1)$ and $Y_n \sim \text{Pois}(\mu_2)$ for $i \in \mathbb{N}$, where $\text{Pois}(\mu)$ denotes the Poisson distribution with parameter μ , for $i \in \{1, 2\}$. For $n \in \mathbb{N}$, the corresponding arithmetic means of X_1, \dots, X_n and Y_1, \dots, Y_n , respectively, are denoted by

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i \quad \text{and} \quad \bar{Y}_n = \frac{1}{n} \sum_{i=1}^n Y_i.$$

In each of the following two tasks, determine the asymptotic variance, i.e. the variance of the corresponding limit distribution, for the sequences of random variables considered in these parts.

(a) For $\mu_1 = 1$, calculate the asymptotic variance σ_1^2 of the sequence $(\sqrt{n}(\bar{X}_n - \mu_1))_{n \geq 1}$. 3 Punkte

Give the value of σ_1^2 . 3 Punkte

(b) For $\mu_2 = 1$, calculate the asymptotic variance σ_2^2 of the sequence $(\sqrt{n}(\sqrt{\bar{Y}_n} - \sqrt{\mu_2}))_{n \geq 1}$. 4 Punkte

Give the value of σ_2^2 . 4 Punkte

Alle Antworten wurden gespeichert

Central Limit Theorem:

$$\lim_{n \rightarrow \infty} \frac{\sqrt{n}(\bar{X}_n - \mu_1)}{\sigma_1} \xrightarrow{d} N(0, 1)$$

$\sigma_1 = 2$

$$\sim \sqrt{n}(\bar{X}_n - \mu_1) \xrightarrow{d} N(0, 4)$$

since for Poisson-Distribution $\text{Var} = \mu$

$\sim \text{Var} = \sigma^2 = 4$

b) delta-method

$\mu_2 = 1$

$$\lim_{n \rightarrow \infty} \frac{\sqrt{n}(\sqrt{\bar{Y}_n} - \sqrt{\mu_2})}{\sigma_2} \xrightarrow{d} N(0, 1)$$

define $g = \sqrt{x}$ and $g' = \frac{1}{2\sqrt{x}} = \frac{1}{2}$

$$\sim \lim_{n \rightarrow \infty} g' \sqrt{n}(\bar{Y}_n - \mu_2) \sim \frac{1}{2} N(0, 1)$$

$$N(0, \frac{1}{4}) = N(0, \frac{1}{4})$$

$\rightarrow \sigma_2^2 = \frac{1}{4}$

Aufgabe 3 10 Punkte

Please provide numbers in the requested precision within each question. The use of different precision is evaluated as wrong.

Consider the following ungrouped data file called `Beetles`, which contains data from one of the first studies using a binary regression model in 1935. The study divided a sample of $n_{\text{tot}} = 481$ adult four beetles to 8 groups of size $n_{ij} = 1, \dots, 8$, with $\sum_{i=1}^8 m_i = n_{\text{tot}}$. The beetles were exposed to gaseous carbon dioxide at 8 distinct dosages (in mg/liter), one for every group. The study observed for the beetles of all groups whether they are alive or dead after a 6 hours exposure, which gives us the corresponding response variable $Y_i \in \{0, 1\}$, where the value 0 denotes the survival of beetle i , $i = 1, \dots, n_{\text{tot}}$. The explanatory variable is the dosage in log-space. Hence, if dosage is the i -th beetle is exposed to, then $x_i = \log_{10}(\text{dose}_i)$, $i = 1, \dots, n_{\text{tot}}$.

1 Punkt

(a) Give the number m_1 of beetles exposed to the dose for which $x_1 = \log_{10}(\text{dose}_1) = 1.691$. Further, give the proportion of deaths among the beetles exposed to this particular dose. 1 Punkt

m_1 (requested precision: whole numbers) 0.5 Punkte

39 Zahl

proportion of deaths (requested precision: 2 digits) 0.5 Punkte

0.10 Zahl

(b) Fit a generalized linear model using the canonical link function for the response Y and treating the explanatory variable as a continuous one. Calculate the sum of squared errors (SSE). 2 Punkte

SSE (requested precision: 2 digits) 2 Punkte

50.14 Zahl

(c) Provide the corresponding AIC and BIC values for the model in (b). 2 Punkte

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AIC (requested precision: 2 digits) 1 Punkt

176.35 Zahl

BIC (requested precision: 2 digits) 1 Punkt

384.7 Zahl

(d) For the model fitted in (b), give the 95% asymptotic (Wald) confidence intervals for the parameter corresponding to the explanatory variable. 2 Punkte

lower bound of CI (requested precision: 3 digits) 1 Punkt

28.107 Zahl

(d) For the model fitted in (b), give the 95% asymptotic (Wald) confidence interval for the parameter corresponding to the explanatory variable. 2 Punkte

lower bound of CI (requested precision: 3 digits) 1 Punkt

28.867

upper bound of CI (requested precision: 3 digits) 1 Punkt

40.328

(e) What is the proportion of correctly classified observations from the model fitted in (b) using 0.5 as threshold probability? 1 Punkt

proportion of correctly classified observations (requested precision: 2 digits) 1 Punkt

0.82 Zahl

(f) Fit a generalized linear model using a probit link. Calculate the sum of squared errors (SSE). Based on the value of SSE, would you prefer the model using a probit link, the model of (d) or both? 2 Punkte

SSE (requested precision: 2 digits) 1 Punkt

59.21

preference based on SSE 1 Punkt

None of the above

change Aufgabe

Aufgabe 4 10 Punkte

The answer of an item of this task is correct (and the corresponding point (points) is (are) granted) only if the correct set of right statements is specified **exactly**.

(a) Choose the statement (statements) that is (are) **true**. 1 Punkt

☐ Logistic regression assumes a linear relationship between the response variable Y and the explanatory variables.

☒ Logistic regression assumes a linear relationship between the logarithm of the odds of the response and the explanatory variables.

☒ If we have a binary response variable, we always have to use logistic regression.

☐ The link function used for obtaining the logistic regression model is the identity link.

Before 6.2.3

☐ The link function used for obtaining the logistic regression model is the log link.

(b) Choose the assumption (assumptions) that is (are) **not** an assumption in the GLM framework where Y is the response variable and X the explanatory variable. 2 Punkte

☒ The response is binary.

☐ The conditional probability density (or mass) function (pdf or pmf) of Y given $X = x$ belongs to the exponential dispersion family.

☒ The conditional probability density (or mass) function (pdf or pmf) of X given $Y = y$ belongs to the exponential dispersion family.

☐ For a random sample of size n , the responses $Y_i, i = 1, \dots, n$, are independent and identically distributed.

☐ The link function links the expectation of the response with the linear predictor.

☐ The link function links the expectation of the response with the linear predictor.

(c) Choose the statement (statements) that is (are) **true**. 2 Punkte

☐ Generalized linear models allow the linear predictor to be non-linear in the parameters β .

☒ Generalized linear models are more sensitive to outliers than linear models.

☐ Generalized linear models can fit complex relationships between the response and the explanatory variables.

☐ Generalized linear models can handle both continuous and categorical data while linear models can just handle one type of them.

☐ In a generalized linear model, the distribution of the error term has to be a normal distribution.

(d) Choose the statement (statements) that is (are) **true** for a GLM. 2 Punkte

☐ For a poisson distributed random response variable, the canonical link is the logit link.

☒ The link function links the expected value of the random response variable to the linear predictor.

☐ The link function transforms the expected value of the random response variable to the natural parameter θ of the exponential dispersion family corresponding to the random response variable.

☐ The link function is used to transform the values of the response variable.

(e) Choose the statement (statements) that is (are) **true** for a GLM. 1 Punkt

☐ The degrees of freedom of a saturated model are always equal to 0.

(f) Consider a simple logistic regression model with parameter vector $\beta = (\beta_1, \beta_2)^T$ and

model matrix $X = \begin{pmatrix} 1 & x_1 \\ 1 & x_2 \\ \vdots & \vdots \\ 1 & x_n \end{pmatrix}$. Choose the statement (statements) that is (are) **true**. 2 Punkte

- ☐ Increasing the explanatory variable by one unit, the odds of success for the response variable will be multiplied by $\exp(\beta_2)$.
- ☐ Increasing the explanatory variable by one unit, the odds of success for the response variable will increase additively by $\exp(\beta_2)$.
- ☐ If $\beta_2 = 0$ the success probability of the response variable is equal to zero.

model matrix $X = \begin{pmatrix} 1 & x_1 \\ \vdots & \vdots \\ 1 & x_n \end{pmatrix}$. Choose the statement (statements) that is (are) **true**. 2 Punkte

- ☒ Increasing the explanatory variable by one unit, the odds of success for the response variable will be multiplied by $\exp(\beta_2)$.
- ☐ Increasing the explanatory variable by one unit, the odds of success for the response variable will increase additively by $\exp(\beta_2)$.
- ☐ If $\beta_2 = 0$ the success probability of the response variable is equal to zero.
- ☒ If $\beta_2 = 0$ the success probability of the response variable is a constant function of the explanatory variable.
- ☐ The median effective level is the point where the success probability of the response variable is maximized.

