

- 7. Consider data on the survival of patients who had undergone surgery for breast cancer. The data consists of a response (survival status after five years) and two predictors (the age of the patient at the time of the operation, and the number of cancerous auxiliary nodes detected):
 - 1. x_1 : Age of patient in years at time of operation (predictor)
 - 2. x_2 : Number of cancerous axillary nodes detected (predictor)
 - 3. Yi: Survival status (response): 0 = the patient survived 5 years or longer; 1 = the patient died within 5 year

Suppose that a logistic regression model, with standardized predictors, correctly fits the data:

$$\eta = \beta_0 + \beta_1 z_1 + \beta_2 z_2 = \log\left(\frac{p}{1-p}\right),$$

where p is the probability of a patient surviving 5 years or longer, and

$$z_j = rac{x_j - mean(x_j)}{sd(x_j)}$$
 for $j = 1, 2$.

Which of the following are correct?

 β_0 represents the mean log odds of surviving 5 years or longer for a person of (sample) mean age, and with the (sample) mean number of cancerous axillary nodes detected.

For a fixed number of cancerous axillary nodes detected, a one standard deviation increase in age increases the odds of survival beyond 5 years by a multiplicative factor of e^{β_1} , on average.

 β_0 represents the mean log odds of surviving 5 years or longer for a person of minimum age and with no cancerous axillary nodes detected.

For a fixed number of cancerous axillary nodes detected, a one standard deviation increase in age increases the log-odds of survival beyond 5 years by β_1 , on average.

For a fixed number of cancerous axillary nodes detected, a one year increase in age increases the log-odds of survival beyond 5 years by β_1 , on average.

- 8. Consider a logistic regression model that uses data to estimate the probability that a client will default on a monthly credit card payment (defaulting on a payment means that the client fails to pay their bill by the deadline for the month in question.)
 - 1. x_1 : credit limit in dollars (**predictor**)
 - 2. x_2 : dollar amount of the bill statement one month prior (predictor)
 - 3. x_3 : dollar amount of the bill statement for two months prior (predictor)
 - 4. x_4 : dollar amount of the payment one month prior (predictor)
 - 5. x_5 : dollar amount of the payment two months prior (predictor)
 - 6. Y_i : default status (response): 0 = the client did not default on the payment for the month in question 5; 1 = the client did default on the payment for the month in question.

Suppose that a logistic regression modelcorrectly fits the data:

$$\eta = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \beta_5 x_5 = \log\left(\frac{p}{1-p}\right),$$

where p is the probability of default.

 $\beta_1/1000$ is the average log-odds of default for a one-thousand dollar increase in credit limit, adjusting for the prior two months' bill statement and payment amounts.

 e^{eta_0} represents the mean odds of defaulting on a payment for a person with a \$0 credit limit, a \$0 bill statement for the last two months, and \$0 in payments for the last two months.

 β_0 represents the mean log-odds of defaulting on a payment for a person with a SO credit limit, a SO bill statement for the last two months, and SO in payments for the last two months.

 e^{eta_1} is the average odds of default for a one-thousand dollar increase in credit limit, adjusting for the prior two months' bill statement and payment amounts.