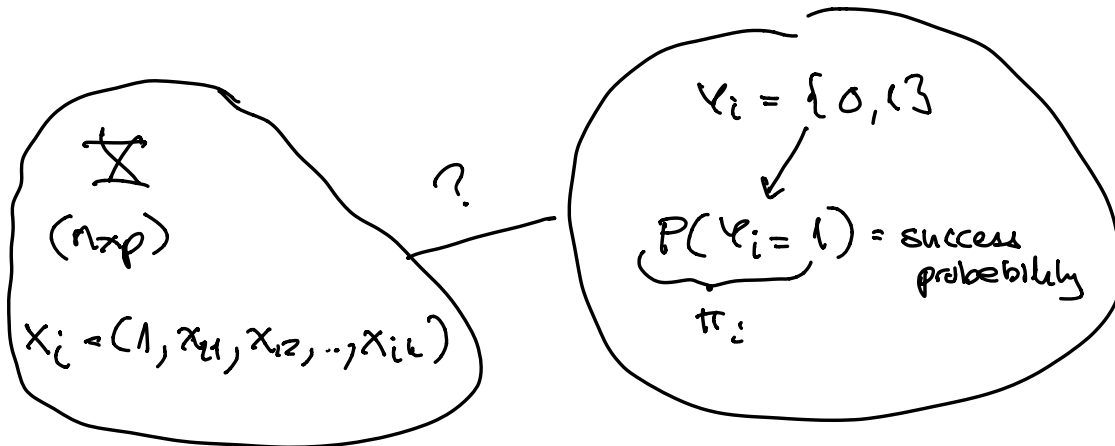


11.09.2017 THA4315 GUT

Module 3: Binary regression (week 1)



Beetle ex: (x_i, Y_i) $i=1, \dots, n$ individual data

Normal model

Binary model

1. $Y_i \sim N(\mu_i, \sigma^2)$

$E(Y_i) = \mu_i$

$Var(Y_i) = \sigma^2$

exponential family distribution

1. $Y_i \sim \text{bin}(n_i=1, \pi_i)$

$E(Y_i) = \pi_i$

$Var(Y_i) = \sigma^2$

2. linear predictor $\eta = x_i^T \beta$

3. $\mu_i = \eta_i$

3. $\mu_i = h(\eta_i) = \text{response function}$

$\eta_i = h^{-1}(\mu_i) = g(\mu_i)$
- link function

The logit model

$$Y_i \sim \text{bin}(n_i=1, \pi_i)$$

$$E(Y_i) = \pi_i, \quad \text{Var}(Y_i) = \pi_i(1-\pi_i)$$

μ_i

$$\mu_i = h(\eta_i) = \frac{\exp(\eta_i)}{1 + \exp(\eta_i)} \quad \left(= \frac{1}{1 + \exp(-\eta_i)} \right)$$

μ_i
 π_i

response function

logistic function

Inverse of response function = link function

$$\eta_i = g(\mu_i) = h^{-1}(\mu_i) : \text{What is this for our model?}$$

$$\mu = \frac{e^{\eta}}{1 + e^{\eta}} \Leftrightarrow (1 + e^{\eta})\mu = e^{\eta}$$

response

$$\mu + \mu e^{\eta} = e^{\eta}$$

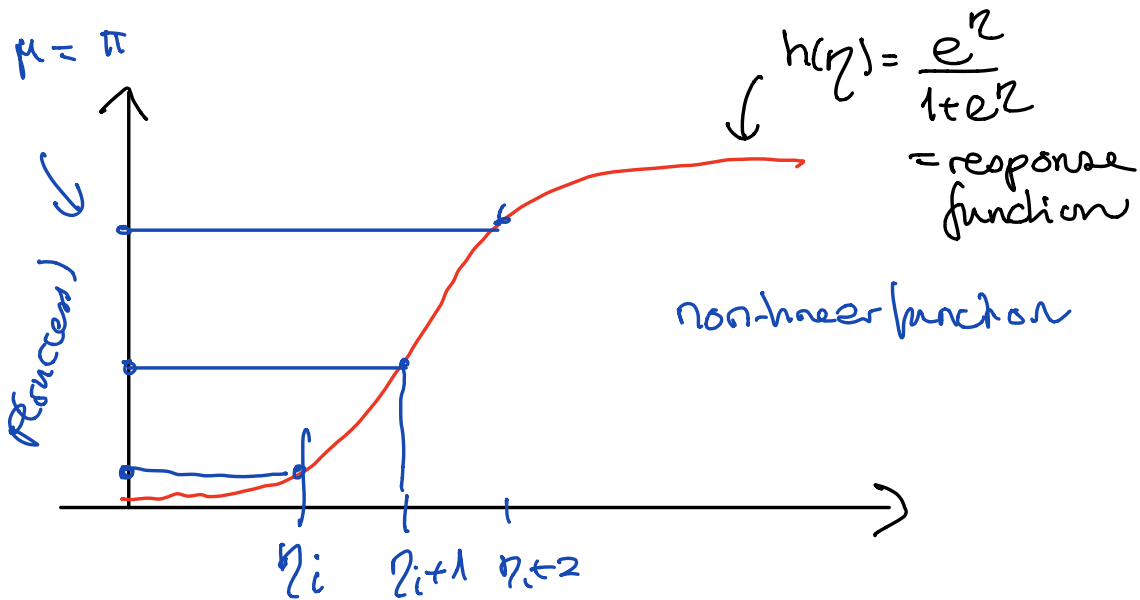
$$\mu = e^{\eta} - \mu e^{\eta} = (1 - \mu)e^{\eta}$$

$$e^{\eta} = \frac{\mu}{1 - \mu}$$

logit link function

$$\eta = \ln\left(\frac{\mu}{1 - \mu}\right) \Rightarrow \eta_i = \ln\left(\frac{\mu_i}{1 - \mu_i}\right)$$

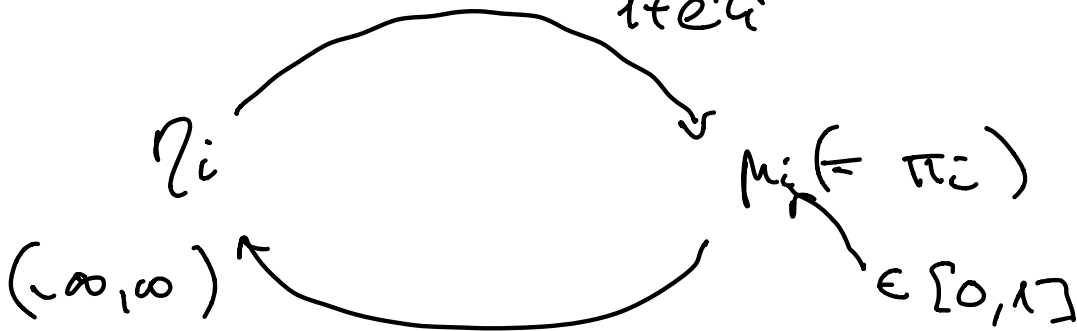
Interpret response function:



$$\eta = \beta_0 + \beta_1 x$$

linear
predictor

response function: $\frac{e^{\eta_i}}{1+e^{\eta_i}}$ ← plot this



link function: $\ln\left(\frac{\pi_i}{1-\pi_i}\right)$

Now - use link function (logit) to aid in interpreting β 's, but first: What is odds?

$$\text{Odds} = \frac{P(Y_i=1)}{P(Y_i=0)} = \frac{\pi_i}{1-\pi_i}$$

$$\pi_i = \frac{1}{5} \Rightarrow \text{odds} = \frac{\frac{1}{5}}{\frac{4}{5}} = \frac{1}{4}, \text{ etc.}$$

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How to interpret a logit model?

$$\pi_i = \frac{\exp(\eta_i)}{1 + \exp(\eta_i)} \Leftrightarrow \eta_i = \ln\left(\frac{\pi_i}{1 - \pi_i}\right)$$

$$\eta_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_k x_{ik}$$

$$\beta_0 + \beta_1 x_{i1} + \dots + \beta_k x_{ik} = \ln\left(\frac{\pi_i}{1 - \pi_i}\right) \quad \text{odds}$$

$$(*) \exp(\beta_0) \cdot \exp(\beta_1 x_{i1}) \cdot \dots \cdot \exp(\beta_k x_{ik}) = \frac{\pi_i}{1 - \pi_i}$$

↑ multiplicative model

So, what if I increase x_{i1} to $(x_{i1} + 1)$ → what happens with $(*)$?

$$\exp(\beta_0) \cdot \underbrace{\exp(\beta_1 x_{i1})}_{\exp(\beta_1 (x_{i1} + 1))} \cdot \exp(\beta_1) \cdot \exp(\beta_2 x_{i2}) \cdot \dots = \frac{\pi_i}{1 - \pi_i} \cdot \exp(\beta_1)$$

⇒ odds multiplied by e^{β_1} .

$\beta_1 = 0 \Rightarrow$ no change $e^0 = 1$

$\beta_1 < 0 \Rightarrow e^{\text{neg}}$ decrease in odds

$\beta_1 > 0 \Rightarrow e^{\text{pos}}$ increase in odds

Ex: infant respiratory disease (see example on module page)

Now: how to interpret the $\hat{\beta}$'s in this example

1) sex Girl: $\text{odds boy} \cdot e^{\hat{\beta}} = \text{odds girl}$ ↑ getting disease
 $\hat{\beta} = -0.3126$ 0.73 $\hat{\beta} < 0 \Rightarrow \text{decrease.}$
 $\exp(\hat{\beta}) = 0.73$ ↑ decrease in odds from boy to girl

food Breast : reference category is bottle
 $\hat{\beta} = -0.6693$
 $\exp(\hat{\beta}) = 0.51$
 $\text{odds bottle} \cdot e^{\hat{\beta}} = \text{odds breast}$
 0.51
 $\Rightarrow \text{decrease in odds from bottle to breast}$

food Suppl
 $\hat{\beta} = -0.1725$: $\text{odds bottle} \cdot e^{\hat{\beta}} = \text{odds Suppl.}$
 $\exp(\hat{\beta}) = 0.84$ 0.84

Now - turn to prediction & interpretation

Six possible covariate patterns:

1: boy & bottle

4: girl & bottle

2: boy & suppl

5: girl & suppl

3: boy & breast

6: girl & breast

$$\text{Prediction: } \hat{\pi} = \frac{\exp(\hat{\beta}_0 + \hat{\beta}_1 x_1 + \hat{\beta}_2 x_2 + \hat{\beta}_3 x_3)}{1 + \exp(\hat{\beta}_0 + \hat{\beta}_1 x_1 + \hat{\beta}_2 x_2 + \hat{\beta}_3 x_3)}$$

x_1 : 0 = boy, 1 = girl

x_2 : 0 = bottle, 1 = suppl

x_3 : 0 = bottle, 1 = breast

Least favourable : $\hat{\pi} = 0.166$ for boys & bottle

Most favourable : $\hat{\pi} = 0.069$ girls & breast

* Mankins \leftarrow bin, logit

\uparrow fit, eval, test, goodness of fit
 \downarrow predic new ≈ 2

Δ football punon n/pzta exh MR

$(H) - (A)$
nair 2x n total

\uparrow
sphi
which & my val