

Ch14 Logistic regression

Y : binary response e.g. 1 or 0
yes or No

"Success" or "Failure"

one predictor: X

~~SLR~~ $Y_i = \beta_0 + \beta_1 X_i + \varepsilon_i \quad E(\varepsilon_i) = 0$

① $\left. \begin{array}{l} \text{If } Y_i = 1, \quad \varepsilon_i = 1 - (\beta_0 + \beta_1 X_i) \\ \text{If } Y_i = 0, \quad \varepsilon_i = 0 - (\beta_0 + \beta_1 X_i) \end{array} \right\} \varepsilon_i \text{ has two values}$

Non-Normality

② restriction on regression function

$$Y_i = \begin{cases} 1 \\ 0 \end{cases} \quad \begin{array}{l} P(Y_i = 1) = \pi_i = p_i \\ P(Y_i = 0) = 1 - \pi_i = 1 - p_i \end{array}$$

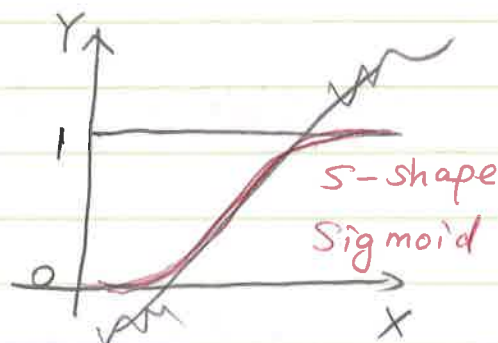
Bernoulli distribution

SLR, $E(Y_i) = \beta_0 + \beta_1 X_i$

$$E(Y_i) = 1 \cdot \underbrace{Pr(Y_i = 1)}_{= \pi_i} + 0 \cdot \cancel{Pr(Y_i = 0)}$$

$$0 \leq \pi_i = \beta_0 + \beta_1 X_i \leq 1$$

$0 \leq \text{Prob} \leq 1$



③ non-constant variance

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

logistic regression work on $\log \frac{\pi_i}{1-\pi_i} = \log \text{Odds}$
 $(-\infty, \infty)$

logit link (probit, log-log)

$$\star \left| \log \frac{\pi_i}{1-\pi_i} = \log \frac{\Pr(Y_i=1)}{1-\Pr(Y_i=1)} = \beta_0 + \beta_1 x_i \right|$$

$$x_i = \begin{cases} 1 & \text{Male} \\ 0 & \text{Female} \end{cases}$$

$$\begin{cases} \log \text{Odds}_{\text{Female}} = \beta_0 \\ \log \text{Odds}_{\text{Male}} = \beta_0 + \beta_1 \end{cases}$$

$$\log \text{Odds}_{\text{male}} - \log \text{Odds}_{\text{female}} = \beta_1$$

$$\log \frac{\text{Odds}_{\text{male}}}{\text{Odds}_{\text{female}}} = \beta_1$$

$$\frac{\text{Odds}_{\text{male}}}{\text{Odds}_{\text{female}}} = \exp(\beta_1) = \exp(0.36) = 1.43$$

The odds that a man is a frequent binge drinker are 1.43 times the odds for a woman.