

LEC. # 8

Nominal multinomial regression

data $(\underline{x}_i, \underbrace{y_{i1}, \dots, y_{iJ}}_{\text{sum to } n_i}) \quad i=1, \dots, n$

on n_i independent trials,
 y_{ij} had the j -th of J
(unordered) outcomes

Binomial when $J=2$
generalizing
to $J>2$

So

$$\Pr\{Y_{i1}=y_{i1}, \dots, Y_{iJ}=y_{iJ}\} = \left(\frac{n_i!}{y_{i1}! \dots y_{iJ}!} \right) p_{i1}^{y_{i1}} \dots p_{iJ}^{y_{iJ}}$$

$$p_{i1} + \dots + p_{iJ} = 1$$

Ch. 5 example

Survey of $n = 944$ voters

Y_i = party identification

"REP.", "DEM.", "IND."

X_i = income, age, education

↑
numeric

↑
categorical
7 categories

So $n_i = 1, i = 1, \dots, 944$
J = 3

Exp. family? GLM?

grey area, since response var. is $J-1$ dimensional

variance is matrix

$$\underline{y}_i = (y_{i1}, y_{i2}, y_{i3})$$

with $y_{i1} + y_{i2} + y_{i3} = 1$

variance function?

Can still think link structure though

multinomial logit model

$$g \begin{pmatrix} p_2 \\ \vdots \\ p_J \end{pmatrix} = \begin{pmatrix} \log \left\{ \frac{p_2}{1 - (p_2 + \dots + p_J)} \right\} \\ \vdots \\ \log \left\{ \frac{p_J}{1 - (p_2 + \dots + p_J)} \right\} \end{pmatrix} = \begin{pmatrix} \underline{\beta}_2^T \\ \vdots \\ \underline{\beta}_J^T \end{pmatrix} \underline{x}$$

vector

and $p_1 = 1 - [p_2 + \dots + p_J]$

$J-1$ rows
cols = # predictors

reduces to logit regression when $J=2$

Interpretation

$$\frac{P_j}{P_i} = e^{\beta_j^T \underline{x}}$$

conditional odds

think constraint
 $\beta_i^T = (0, \dots, 0)$

$$\frac{P_j/P_i}{P_k/P_i}$$

$$\frac{P_j}{P_k} = e^{(\beta_j - \beta_k)^T \underline{x}}$$

$$\text{Odds}(Y=j | Y \in \{j, k\}) = \frac{P_j}{P_j + P_k}$$

$$(P_1, \dots, P_J) = \frac{1}{(1 + \sum_{j=2}^J e^{\beta_j^T \underline{x}})} (1, e^{\beta_2^T \underline{x}}, \dots, e^{\beta_J^T \underline{x}})$$

inverting the link

Note: typically want
 $\underline{x} = (1, \text{predictors})$

$$(P_1, \dots, P_J) = \frac{1}{1 + \sum_{j=2}^J e^{\beta_j^T \underline{x}}} (1, e^{\beta_2^T \underline{x}}, \dots, e^{\beta_J^T \underline{x}})$$

first col. of B are intercept params.

so predictors = 0

Ex.

recall
"DEM"
is reference
level

$$\hat{\beta} = \begin{matrix} & \text{Intercept} & \text{Income (\$1000)} \\ \begin{matrix} \text{"IND"} \\ \text{"REP"} \end{matrix} & \begin{pmatrix} -1.18 \\ -0.95 \end{pmatrix} & \begin{pmatrix} 0.0161 \\ 0.0176 \end{pmatrix} \end{matrix}$$

$$P|INC=0 \propto (e^0, e^{-1.18}, e^{-0.95})$$

hypothetical = (0.59, 0.18, 0.23)

"D" "I" "R"

Cond. OR associated with
\$1000 increase in income?

$$\frac{P_2(z+1000)}{P_1(z+1000)}$$

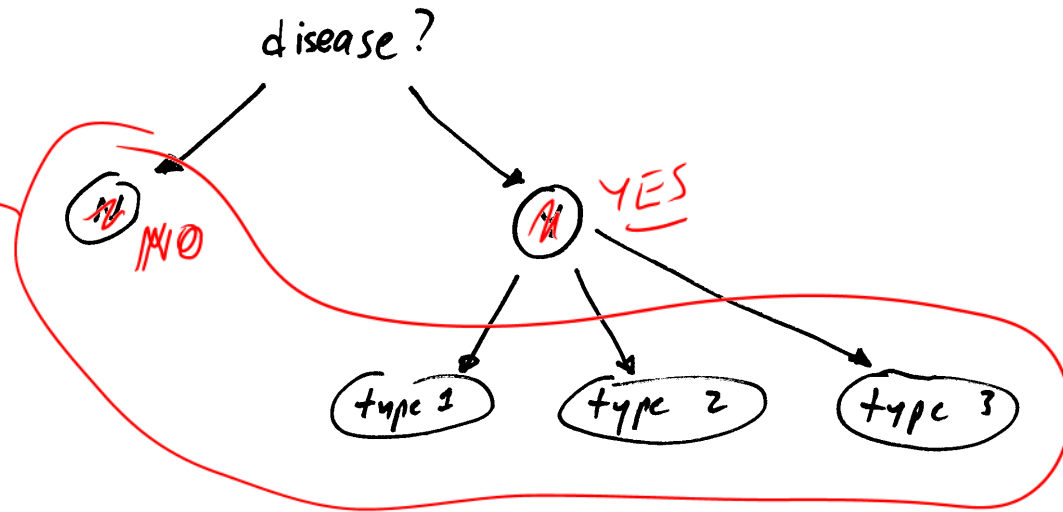
$$\frac{P_2(z)}{P_1(z)}$$

$$e^{.0161} \approx 1.0162$$

$$e^{.017} \approx 1.0178$$

Hierarchical / Nested

4
categories



Strategy - separately fit

- Binomial model for DISEASE
- Multinomial model for TYPE/DISEASE=YES

Statistical justification? ✓

Start with multinomial model

$$\text{lik.} \propto \prod_{i=1}^n p_1^{y_{i1}} p_2^{y_{i2}} p_3^{y_{i3}} p_4^{y_{i4}}$$

reparameterize $q = \Pr\{\text{Dis.} = Y\} = p_2 + p_3 + p_4$

$$r_i = \Pr\{\text{Type} = i \mid \text{Dis.} = Y\} = \frac{p_{i+1}}{p_2 + p_3 + p_4}$$

So lik. $\propto \prod_{i=1}^n (1-q)^{y_{i1}} (qr_1)^{y_{i2}} (qr_2)^{y_{i3}} (qr_3)^{y_{i4}}$

$$\propto \left\{ \prod_{i=1}^n (1-q)^{y_{i1}} q^{y_{i2} + y_{i3} + y_{i4}} \right\} \left\{ \prod_{i=1}^n r_1^{y_{i2}} r_2^{y_{i3}} r_3^{y_{i4}} \right\}$$

depends on q only

depends on (r_1, r_2, r_3) only
 factors -
 can fit
 separate models

$$1-q =$$

$$\Pr\{\text{Dis.} = N_0\}$$

stats
 matches
 intuition

Fundamentally different modelling assumptions by choosing different parameterization for "linking"

e.g. nested model says

$$\log\left(\frac{p_2 + p_3 + p_4}{p_1}\right) \text{ linear in } \underline{x}$$

disease (above numerator) *no disease* (below denominator)

regular multinomial model

says

$$\log\left(\frac{p_2 + p_3 + p_4}{p_1}\right) = \log\left\{ e^{\beta_2^T \underline{x}} + e^{\beta_3^T \underline{x}} + e^{\beta_4^T \underline{x}} \right\}$$

not linear in \underline{x}