

## Travel Data Example

4/5/22

mode ~ hinc + psize

Baseline:  $\gamma = \text{air}$

$x_1 = \text{household income } (\$/\text{1000})$

$x_2 = \text{party size}$

Fitted model:

$$\log \left( \frac{\hat{\pi}_T}{\hat{\pi}_A} \right) = \exp [1.550 - 0.0609 x_1 + 0.2907 x_2]$$

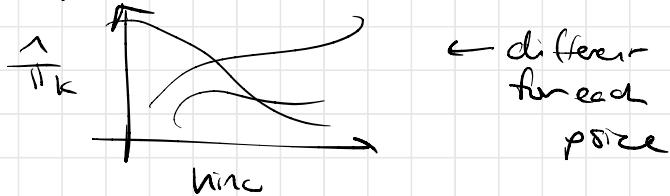
$$\log \left( \frac{\hat{\pi}_B}{\hat{\pi}_A} \right) = 1.034 - 0.0339 x_1 - 0.3397 x_2$$

$$\log \left( \frac{\hat{\pi}_C}{\hat{\pi}_A} \right) = -0.944 - 0.00354 x_1 + 0.6006 x_2$$

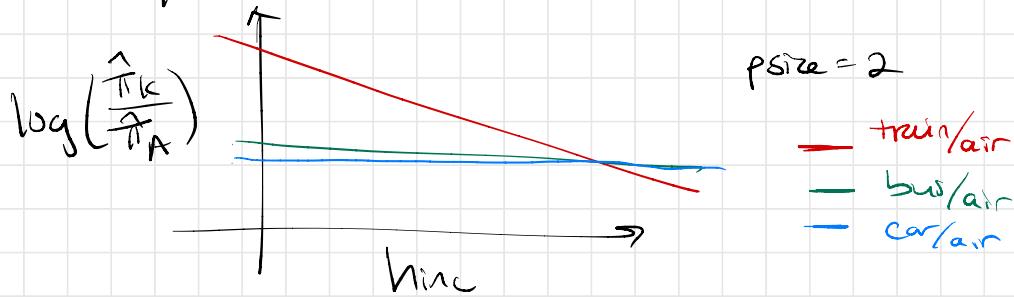
What about estimated conditional odds of train travel  
vs. car travel?

$$\begin{aligned} \log \left( \frac{\hat{\pi}_T / \hat{\pi}_A}{\hat{\pi}_C / \hat{\pi}_A} \right) &= \log \left( \frac{\hat{\pi}_T}{\hat{\pi}_C} \right) - \log \left( \frac{\hat{\pi}_A}{\hat{\pi}_A} \right) \\ &= (1.550 + 0.944) + (-0.0609 + 0.00354) x_1 \\ &\quad + (0.2907 - 0.6006) x_2 \\ &= 2.49 - 0.058 x_1 - 0.31 x_2 \end{aligned}$$

# Visualizing - plot of probabilities



Plot log cond. odds



Who interaction → intercepts change w/psize  
but not slopes

## Linear combinations of coeffs.

95% CI for RRR of train to air for a \$1000 increase in hinc (for parties of size 2)

$$\frac{\text{TR}}{\text{TA}}$$

$x_1 = \text{household income ($1000)}$   
 $x_2 = \text{party size}$

$$\log\left(\frac{\text{TR}}{\text{TA}}\right) = \beta_{01} + \beta_{11}x_1 + \beta_{21}x_2 + \beta_{31}x_1x_2$$

$$x_1 \uparrow, \quad \begin{cases} \beta_{01} + \beta_{11}x_1 + \beta_{21}(2) + \beta_{31}x_1(2) \\ = \beta_{01} + \beta_{21}(2) + (\beta_{11} + 2\beta_{31})x_1 \end{cases}$$

$$95\% \text{ CI for } e^{\beta_{11} + 2\beta_{31}} : (0.926, 0.973)$$

-7.4% - 2.7%

Interpret:

We are 95% confident that the true change in relative risk ratio of travel by train to air for a \$1000 increase in household income is between a 2.7% to 7.4% decrease, among parties of size two.

$$\text{CI for } \frac{\text{TR}}{\text{TA}}$$

We are 95% confident that the relative risk ratio of travel by train to air / the conditional odds of travel by train compared to air decreases by between 2.7% to 7.4% for each \$1000 increase in household income when the party size is 2 individuals.

## Section 6.2 — Cumulative Logit Models for ordinal response

$$Y_i = \begin{cases} 1 & \text{Prob.} \\ 2 & \pi_1 \\ 3 & \pi_2 \\ \vdots & \vdots \\ J & \pi_J \end{cases}$$

← labels reflect natural ordering

e.g. low, med, high  
— Likert scale

$$\pi_{ik} = P(Y_i = k)$$

A cumulative probability is the probability of falling at or below a particular category:

$$P(Y_i \leq k) = \pi_1 + \pi_2 + \dots + \pi_k$$

- Notes:
- ①  $P(Y_i \leq J) = 1$  ← reflects ordering
  - ②  $P(Y_i \leq 1) \leq P(Y_i \leq 2) \leq \dots \leq P(Y_i \leq J)$

For  $k = 1, 2, \dots, J-1$

$$\text{logit}(P(Y_i \leq k)) = \beta_{0k} + \beta_{1k} x_1 + \dots + \beta_{mk} x_m$$

$$\text{"cumulative logit"} = \log \left( \frac{P(Y_i \leq k)}{1 - P(Y_i \leq k)} \right)$$

$$= \log \left( \frac{P(Y_i \leq k)}{P(Y_i > k)} \right) = \log \left( \frac{\pi_1 + \pi_2 + \dots + \pi_k}{\pi_{k+1} + \dots + \pi_J} \right)$$